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# POLISH PHONEME STATISTICS OBTAINED ON LARGE SET OF WRITTEN TEXTS

The phonetical statistics were collected from several Polish corpora. The paper is a summary of the data which are phoneme n-grams and some phenomena in the statistics. Triphone statistics apply context-dependent speech units which have an important role in speech recognition systems and were never calculated for a large set of Polish written texts. The standard phonetic alphabet for Polish, SAMPA, and methods of providing phonetic transcriptions are described.

Keywords: NLP, triphone statistics, speech processing, Polish

# STATYSTYKI POLSKICH FONEMÓW UZYSKANE Z DUŻYCH ZBIORÓW TEKSTÓW

W niniejszej pracy zaprezentowano opis statystyk głosek języka polskiego zebranych z dużej liczby tekstów. Triady głosek pełnia istotną rolę w rozpoznawaniu mowy. Omówione obserwacje dotyczące zebranych statystyk i przedstawiono listy najpopularniejszych elementów.

Słowa kluczowe: przetwarzanie języka naturalnego, statystyki głosek, przetwarzanie mowy

## 1. Introduction

The authors uses the Cyfronet, high performance computers to process linguistic data in aim to construct the Polish language models. The results will be applied to a large vocabulary continuous speech recognition system (LVCSR). Natural language processing (NLP) faces problems of data sparsity very often. The quality of language models is strongly dependent on the amount of text corpora available during the training. This is why, there is a trade-off of quality and time spent on calculations. The high performance computers facilitate obtaining the linguistic rules from the huge amount of texts.

Statistical linguistics at the word and sentence level were under considerations for several languages [1, 2]. However, similar research on phonemes is rare [3, 4, 5]. The frequency of phonetic units appearance is an important topic itself for every

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language. It can also be used in several speech processing applications, for example modelling in LVCSR or coding and compression. Models of triphones which are not present in a training corpus of a speech recogniser can be prepared using phonetic decision trees [6]. The list of possible triphones has to be provided for a particular language along with phonemes' categorisation. The triphone statistics can also be used to generate hypotheses used in recognition of out-of-dictionary words including names and addresses.

We have already presented some similar statistics [7], which were collected from around 10 000 000 words of mainly spoken language. Data collected from a few much larger corpora: Rzeczpospolita corpus (containing articles from a well known in Poland, everyday newspaper of quality and type like Times or Guardian), literature corpus and Internet encyclopedia corpus are presented in this work combined statistical. The presented statistics are the biggest and most representative statistics of phonemes for Polish. They were collected from over 250 000 000 words.

### 2. Description of a problem solution

The problem is to find triphone statistics for Polish language. Our first attempt to this task was already published [7]. The task was conducted on a corpus containing Parliament transcriptions mainly (around 50 megabytes of text). It was repeated on Mars, a Cyfronet computer cluster, for data of around 2 gigabytes.

Context-dependent modelling can significantly improve speech recognition quality. Each phoneme varies slightly depending on its context, namely neighbouring phonemes due to a natural phenomena of coarticulation. It means that there are no clear boundaries between phonemes and they overlap each other. It results in interference of acoustical properties. Speech recognisers based on triphone models rather than phoneme ones are much more complex but give better results [9]. Let us present examples of different ways of transcribing word *above*. Phoneme model is  $ax \ b \ ah \ v$ while the triphone one is \*- $ax+b \ ax-b+ah \ b-ah+v \ ah-v+*$ . In case a specific triphone is not present, it can be replaced by a phonetically similar triphone (phonemes of the same phonetic group interfere in similar way with their neighbours) using phonetic decision trees [6] or diphones (applying only left or right context) [10].

### 3. Methods, software and hardware

Sophisticated rules and methods are necessary to obtain the phonetic information from an orthographic text-data. Simplifications could cause errors [11]. Transcription of text into phonetic data was applied first by PolPhone [8]. The extended SAMPA phonetic alphabet was applied with 39 symbols (plus space) and pronunciation rules for cities Poznań and Kraków. We used our own digit symbols corresponding to SAM-PA symbols, instead of typical ones, to distinguish phonemes easier while analysing received phonetic transcriptions.

SAMPA	example	transcr.	occurr.	%	% [5]
#		#	283 296 436	15.256	4.7
a	pat	$\operatorname{pat}$	$151 \ 160 \ 947$	8.141	9.7
е	test	$\operatorname{test}$	$146 \ 364 \ 208$	7.882	10.6
0	pot	$\operatorname{pot}$	$141 \ 975 \ 325$	7.646	8.0
$\mathbf{t}$	test	$\operatorname{test}$	68 851 605	3.708	4.8
r	ryk	rIk	$68\ 797\ 073$	3.705	3.2
n	nasz	naS	$68 \ 056 \ 439$	3.665	4.0
i	PIT	$\operatorname{pit}$	$67 \ 212 \ 728$	3.620	3.4
j	jak	jak	$61\ 265\ 911$	3.299	4.4
Ι	typ	$_{\mathrm{tIp}}$	$58 \ 930 \ 672$	3.174	3.8
v	wilk	vilk	$58\ 247\ 951$	3.137	2.9
s	syk	$_{ m sIk}$	$54 \ 359 \ 454$	2.927	2.8
u	puk	puk	$51 \ 503 \ 621$	2.774	2.8
р	pik	pik	$51\ 228\ 649$	2.759	3.0
m	mysz	mIS	$48\ 760\ 010$	2.626	3.2
k	kit	$_{\rm kit}$	44 892 420	2.418	2.5
d	dym	dIm	$44 \ 406 \ 412$	2.391	2.1
1	luk	luk	40 189 121	2.164	1.9
n'	koń	kon'	$34\ 092\ 610$	1.84	2.4
z	zbir	zbir	$30 \ 924 \ 282$	1.665	1.5
w	łyk	wIk	$30\ 194\ 178$	1.626	1.8
f	fan	fan	$25 \ 308 \ 167$	1.363	1.3
g	gen	gen	$24 \ 910 \ 462$	1.341	1.3
$t^s$	cyk	$t^sIk$	$24\ 789\ 080$	1.335	1.2
b	bit	bit	$24 \ 212 \ 663$	1.304	1.5
x	hymn	xImn	$21 \ 407 \ 209$	1.153	1.0
$\mathbf{S}$	szyk	SIk	$20\ 756\ 164$	1.118	1.9
$\mathbf{s}'$	świt	s'vit	$17\ 220\ 321$	0.927	1.6
Z	żyto	ZIto	$16\ 409\ 930$	0.884	1.3
$t^S$	czyn	$t^SIn$	$15\ 429\ 711$	0.831	1.2
$t^s'$	ćma	$t^s'ma$	$11 \ 945 \ 381$	0.643	1.2
$ m w\sim$	ciąża	ts'ow $\sim$ Za	$10\ 814\ 216$	0.582	0.6
с	kiedy	cjedy	$10\ 581\ 296$	0.570	0.7
d^z'	dźwig	d^z'vik	$9 \ 995 \ 596$	0.538	0.7
Ν	pęk	$\operatorname{peNk}$	$4\ 880\ 260$	0.262	0.1
d^z	dzwoń	d^zvon'	$4\ 212\ 857$	0.227	0.2
J	giełda	Jjewda	$3\ 680\ 888$	0.198	0.1
z'	źle	z'le	$3 \ 390 \ 372$	0.183	0.2
j~	więź	vjej∼s'	$1 \ 527 \ 778$	0.082	0.1
$d^{2}$	dżem	$d^{2}em$	693 838	0.037	0.1

Table 1Phonemes in Polish (SAMPA [8])

Stream editor (SED) was applied to change original phoneme transcriptions into digits with the following script:

s/##/#/g	$s/w\sim/2/g$	$s/d^2/6/g$
$s/t^s'/8/g$	s/s'/5/g	$\rm s/t^S/0/g$
$s/d^z'/X/g$	s/z'/4/g	$\rm s/d^2/9/g$
$s/j\sim/1/g$	$s/t^s/7/g$	s/n'/3/g.

Statistics can now be simply collected by counting the number of occurrences of each phoneme, phoneme pair, and phoneme triple in an analysed text, where each phoneme is just a symbol (single letter or a digit). Matlab was used to analyse the phonetic transcription of the text corpora. The calculations were conducted on Mars in Cyfronet, Krakow. We analysed more than 2 gigabytes of data. Text data for Polish are still being collected and will be included in the statistics in the future.

Mars is a cluster for calculations with following specification: IBM Blade Center HS21 – 112 Intel Dual-core processors, 8 GB RAM/core, 5 TB disk storage and 1192 Gflops. It operates using Red Hat Linux. Mars uses Portable Batch System (PBS) to queue tasks and split calculation power to optimise times for all users. A user have to declare expected time of every task. In example, a short time is up to 24 hours of calculations and a long one is up to 300 hours. Tasks can be submitted by simple commands with scripts and the cluster starts particular tasks when calculation resources are available. One process needs around 100 hours to analyse 45 megabytes text file.

#### 3.1. Grapheme to phoneme transcription

Two main approaches are used for the automatic transcription of texts into phonemic forms. The classical approach is based on phonetic grammatical rules specified by human [12] or machine learning process [13]. The second solution utilises graphemicphonetic dictionaries. Both methods were used in PolPhone to cover typical and exceptional transcriptions. Polish phonetic transcription rules are relatively easy to formalise because of their regularity.

The necessity of investigating large text corpus pointed to the use of the Polish phonetic transcription system PolPhone [14, 8]. In this system, strings of Polish characters are converted into their phonetic SAMPA representations. Extended SAMPA (Table 1) is used, to deal with nuances of Polish phonetic system. The transcription process is performed by a table-based system, which implements the rules of transcription. Matrix  $T \in S^{m \times n}$  is a transcription table, where S is a set of strings and the cells meet the requirements listed precisely in [8]. The first element  $t_{1,1}$  of each table contains currently processed character of the input string. For every character (or character substring) one table is defined. The first column of each table  $\{t_{i,1}\}_{i=1}^{m}$ contains all possible character strings that could precede currently transcribed character. The first row  $\{t_{1,j}\}_{j=1}^{n}$  contains all possible character strings that can proceed a currently transcribed character. All possible phonetic transcription results are stored in the remaining cells  $\{t_{i,j}\}_{i=2,j=2}^{m,n}$ . A particular element  $t_{i,j}$  is chosen as a transcription result, if  $t_{i,1}$  matches the substring preceding  $t_{1,1}$  and  $t_{1,j}$  matches the substring proceeding  $t_{1,1}$ . This basic scheme is extended to cover overlapping phonetic contexts. If more then one result is possible, then longer context is chosen for transcription, which increases its accuracy. Exceptions are handled by additional tables in the similar manner.

Specific transcription rules were designed by a human expert in an iterative process of testing and updating rules. Text corpora used in design process consisted of various sample texts (newspaper articles) and a few thousand words and phrases including special cases and exceptions.

#### 3.2. Corpora used

Several newspaper articles in Polish were used as input data in our experiment. They are from Rzeczpospolita newspaper from years 1993–2002. They cover mainly political and economic issues, so they contain quite many names and places including foreign ones, what may influence the results slightly. In example, q appeared once, even though it does not exist in Polish. In total, 879 megabytes of text, which corresponds to around 110 000 000 words, were included in the process.

Several hundreds of thousands of Internet articles in Polish made another corpus. They are all from a high quality website, where all content is reviewed and controlled by moderators. They are of encyclopedia type, so they also contain many names including foreign ones. In total, 754 megabytes (around 94 000 000 words) were included in the process.

The third corpus consists of several literature books in Polish. Some of them are translations from other languages, so they also contain foreign words. The corpus includes 490 megabytes (around 61 000 000 words) of text.

### 4. Results

The total number of around 1856 900 000 phonemes were analysed. They are grouped into 40 categories (including space). Actually, one more, namely q, was detected, which appeared in a foreign name. Since q is not a part of the Polish alphabet, it was not included in the phoneme distribution presented in Table 1. Space (noted as #) frequency was 15.26 %. An average number of phonemes in words is 6.6 including one space. Exactly 1271 different diphones (Fig. 1 and Table 2) for 1560 possible combinations were found, which constitutes 81%.

21 961 different triphones (see Table 3) were detected. Combinations like  $*#^*$ , where \* is any phoneme and # is a space were removed. These triples should not be considered as triphones because the first and the second \* are in two different words. The list of the most common triphones is presented in Table 3. Assuming 40 different phonemes (including space) and subtracting mentioned  $*#^*$  combinations, there are 62 479 possible triples. We found 21 961 different triphones. It leads to a conclusion that around 35% of possible triples were detected as triphones, the very most of them at least 10 times.

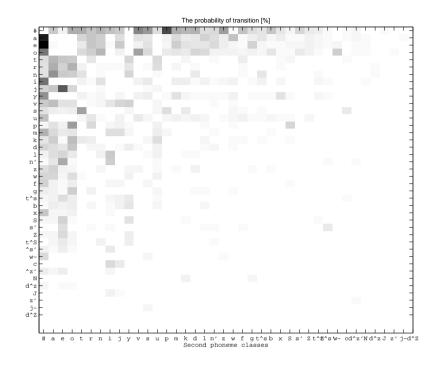


Fig. 1. Frequency of diphones in Polish (each phoneme separately)

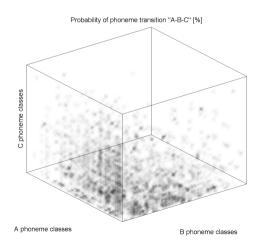


Fig. 2. Space of triphones in Polish

diphone	no. of occurr.	%	diphone	no. of occurr.	%
e#	43 557 832	2.346	on	12 854 255	0.692
a#	$38 \ 690 \ 469$	2.084	#k	$12 \ 529 \ 124$	0.675
#p	$31 \ 014 \ 275$	1.671	ta	$12 \ 449 \ 178$	0.671
je	$28 \ 499 \ 593$	1.535	#n	$12 \ 316 \ 393$	0.663
i#	$24 \ 271 \ 474$	1.307	va	$11 \ 413 \ 878$	0.615
o#	$23 \ 552 \ 591$	1.269	ko	$11\ 168\ 294$	0.602
#v	$20\ 678\ 007$	1.114	#i	$10 \ 515 \ 253$	0.566
y#	$19\ 018\ 563$	1.024	aw	$10 \ 514 \ 514$	0.566
na	$18 \ 384 \ 584$	0.990	u#	$10 \ 379 \ 234$	0.559
#s	$17 \ 321 \ 614$	0.933	#f	$10\ 265\ 162$	0.553
ро	$16\ 870\ 118$	0.909	#b	$10\ 167\ 482$	0.548
#z	$16 \ 619 \ 556$	0.895	#r	$10\ 137\ 129$	0.546
ov	$16\ 206\ 857$	0.873	ja	$10\ 097\ 444$	0.544
$\mathbf{st}$	$15 \ 895 \ 694$	0.856	ar	$9\ 818\ 127$	0.529
n'e	$14 \ 851 \ 771$	0.800	x#	$9\ 811\ 211$	0.528
#o	$14 \ 104 \ 742$	0.760	do	$9\ 779\ 666$	0.527
#t	$13 \ 910 \ 147$	0.749	er	$9\ 724\ 692$	0.524
ra	$13\ 713\ 928$	0.739	te	$9\ 618\ 998$	0.518
#m	$13 \ 657 \ 073$	0.736	#j	$9 \ 398 \ 210$	0.506
ro	$13 \ 597 \ 891$	0.732	v#	$9\ 251\ 288$	0.498
#d	$13 \ 103 \ 398$	0.706	#a	$9\ 143\ 021$	0.492
m#	$12 \ 968 \ 346$	0.698	to	$9\ 043\ 529$	0.487

Table 2Most common Polish diphones

Young [9], estimates that in English, 60–70% of possible triples exist as triphones. However, in his estimation there is no space between words, what changes the distribution a lot. Some triphones may not occur inside words but may occur at combinations of an end of one word and the beginning of another. We started to calculate such statistics without an empty space as the next step of our research. It is also expected that there are different numbers of triphones for different languages. Some values are similar to statistics given by Jassem a few decades ago and reprinted in [5]. We applied computer clusters so our statistics were calculated for much more data and they are more representative.

Fig. 1 shows some symmetry but the probability of diphone  $\alpha\beta$  is usually different than probability of  $\beta\alpha$ . The mentioned quasi symmetry results from the fact that high values of  $\alpha$  probability and (or)  $\beta$  probability often gives high probability of products  $\alpha\beta$  and  $\beta\alpha$  as well. Similar effects can be observed for triphones. Data presented in this paper illustrate the well-known fact that probabilities of triphones (see Table 3) cannot be calculated from the diphone probabilities (see Table 2). The conditional probabilities between diphones have to be known.

triphone	no. of occurr.	%	triphone	no. of occurr.	%
#po	12 531 515	0.675	wa#	3 262 204	0.176
#na	$9\ 587\ 483$	0.516	do#	$3\ 210\ 532$	0.173
n'e#	$9\ 178\ 080$	0.494	#ma	$3\ 209\ 675$	0.173
na#	8 588 806	0.463	jon	$3\ 082\ 879$	0.166
ow~#	6778259	0.365	e#z	$3\ 054\ 967$	0.165
#do	$6\ 751\ 495$	0.364	a#v	$3\ 028\ 787$	0.163
#za	$6\ 429\ 379$	0.346	#z#	$2 \ 928 \ 164$	0.158
ej#	$6 \ 390 \ 911$	0.344	ka#	2 871 230	0.155
je#	$6 \ 388 \ 032$	0.344	#sp	2 818 515	0.152
#pS	$6\ 173\ 458$	0.333	ont^s	2754934	0.148
go#	5 990 895	0.323	e#s	2 737 210	0.147
#i#	5 945 409	0.320	i#p	$2\ 725\ 414$	0.147
ego	5 742 711	0.309	o#p	2 719 121	0.146
ova	$5\ 560\ 749$	0.300	#Ze	2 701 194	0.145
vje	$5\ 433\ 154$	0.293	#ja	$2\ 670\ 034$	0.144
#v#	$5\ 317\ 078$	0.286	ta#	$2\ 618\ 595$	0.141
#je	$5\ 311\ 716$	0.286	ent	$2\ 612\ 166$	0.141
#n'e	$5\ 292\ 103$	0.285	#to	2 567 269	0.138
sta	4 983 295	0.268	to#	$2\ 557\ 630$	0.138
#s'e	4 861 117	0.262	pro	$2\ 548\ 979$	0.137
yx#	4 858 960	0.262	pra	2 539 424	0.137
#vy	4763697	0.257	#pa	2 503 153	0.135
s'e#	$4\ 746\ 280$	0.256	#re	2 502 443	0.135
pSe	$4\ 728\ 565$	0.255	ost	$2 \ 490 \ 304$	0.134
e#p	4 727 840	0.255	#ty	$2\ 452\ 830$	0.132
#f#	4 660 745	0.251	t^se#	$2\ 436\ 864$	0.131
em#	4 514 478	0.243	#mj	$2 \ 397 \ 741$	0.129
$\# \mathrm{pr}$	$4 \ 428 \ 341$	0.239	ku#	$2 \ 383 \ 231$	0.128
#ko	$4\ 216\ 459$	0.227	e#m	$2 \ 379 \ 510$	0.128
a#p	$4\ 155\ 732$	0.224	ja#	$2 \ 353 \ 638$	0.127
ci#	$3 \ 965 \ 693$	0.214	e#o	$2 \ 343 \ 622$	0.126
ne#	$3 \ 958 \ 262$	0.213	a#s	$2 \ 336 \ 272$	0.126
cje	$3 \ 916 \ 595$	0.211	#vj	$2 \ 329 \ 962$	0.125
n'a#	$3\ 888\ 279$	0.209	#mo	$2 \ 320 \ 091$	0.125
#ro	$3\ 785\ 754$	0.204	nyx	$2 \ 299 \ 719$	0.124
mje	$3\ 760\ 340$	0.203	os't^s'	$2 \ 295 \ 365$	0.124
#st	$3\ 745\ 320$	0.202	ovy	$2 \ 284 \ 782$	0.123
aw#	3 596 680	0.194	sci	$2 \ 282 \ 887$	0.123
ny#	3 580 425	0.193	ove	$2 \ 262 \ 277$	0.122
#te	$3\ 449\ 304$	0.186	li#	$2\ 255\ 403$	0.121
e#v	$3 \ 313 \ 798$	0.178	ovj	$2 \ 251 \ 294$	0.121
$\mathrm{Ze}\#$	$3 \ 309 \ 352$	0.178	mi#	$2 \ 243 \ 432$	0.121
ym#	3 300 273	0.178	uv#	$2 \ 236 \ 507$	0.120

Table 3Most common Polish triphones

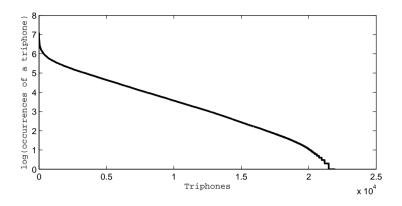


Fig. 3. Phoneme occurrences distribution

Besides the frequency of triphone occurrence, we are also interested in distributions of their frequencies. This is presented in logarithmic scale in Fig. 3. We received another distribution than in the previous experiment [7] because larger number of words were analysed. We have found around 500 triphones which occurred once and around 300 which occurred two or three times. Then every occurrence up to 10 happened for 100 to 150 triphones. It supports a hypothesis that one can reach a situation, when new triphones do not appear and a distribution of occurrences is changing as a result of more data being analysed. Some threshold can be set and the rarliest triphones can be removed as errors caused by unusual Polish word combinations, acronyms, slang and other variations of dictionary words, onomatopoeic words, foreign words, errors in phonisation and typographical errors in the text corpus.

Entropy:

$$H = -\sum_{i=1}^{40} p(i) \log_2 p(i), \tag{1}$$

where p(i) is a probability of a particular phoneme, is used as a measure of the disorder of a linguistic system. It describes how many bits in average are needed to describe phonemes. According to Jassem in [5] entropy for Polish is 4.7506 bits/phoneme. From our calculations entropy for phonemes is 4.6335, for diphones 8.3782 and 11.5801 for triphones.

## 5. Conclusions

250 000 000 words from different corpora: newspaper articles, Internet and literature were analysed. Statistics of Polish phonemes, diphones and triphones were created. They are not fully complete, but the corpora were large enough, that they can be successfully applied in NLP applications and speech processing. The collected statistics are the biggest for Polish of this type of linguistic computational knowledge. Polish is

one of most common Slavic languages. It has several different phonemes than English and the statistics of phonemes are also different.

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