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# Compression principle and Zipf's law of brevity in infochemical communication

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## Abstract

Compression has been presented as a general principle of animal communication. Zipf's law of brevity is a manifestation of this postulate and can be generalised as the tendency of more frequent communicative elements to be shorter. Previous works supported this claim, showing evidence of Zipf's law of brevity in animal acoustical communication and human language. However, a significant part of the communicative effort in biological systems is carried out in other transmission channels, such as those based on infochemicals. To fill this gap, we seek, for the first time, shreds of evidence of this principle in infochemical communication by analysing the statistical tendency of more frequent infochemicals to be chemically shorter and lighter. We analyse data from the largest and most comprehensive open-access infochemical database known as Pherobase, recovering Zipf's law of brevity in interspecific communication (allelochemicals) but not in intraspecific communication (pheromones). Moreover, these results are robust even when addressing different magnitudes of study or mathematical approaches. Therefore, different dynamics from the compression principle would dominate intraspecific chemical communication, defying the universality of Zipf's law of brevity. To conclude, we discuss the exception found for pheromones in the light of other potential communicative paradigms such as pressures on successful communication or the Handicap principle.

**Keywords:** compression, infochemicals, information theory, language, communication.

## 1 Introduction

Chemical communication plays a paramount role in the complex dynamic of ecological systems and has governed life from the very beginning, from DNA replication to the internal communication mechanisms of living beings [1]. It is crucial for understanding the behaviour of organisms [2] and has shaped ecological communities [3, 4]. One form of chemical communication is infochemicals, substances emitted outside of the body and carrying information to interact with other individuals from the same or different species. Infochemicals play a relevant role in complex ecosystems and have recently been the subject of intense research in ecology [5, 6], evolution [7] and environmental

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27 [8] or agricultural sciences [9], among other scientific fields.

28

29 Infochemicals have evolved under communication networks' selective pressures [10], from non-  
 30 communicative compounds with a phylogenetic pattern [11, 12], to aggregates specialised in effi-  
 31 ciently accomplishing the transmission of biological information. Each infochemical has been clas-  
 32 sically assorted considering five adaptative functions depicted in Figure 1: pheromones, involved in  
 33 intraspecific communication; attractants, that cause aggregation of individuals; allomones, when  
 34 they benefit the sender; kairomones when the beneficial is the receiver; and finally synomones,  
 35 benefiting both signaller and receiver in mutualistic interactions. Definitely, in a relevant dis-  
 36 tinction here, attractants, allomones, kairomones and synomones are allelochemicals that mediate  
 37 interspecies communications, while pheromones intercede in intraspecific communication [13].

38

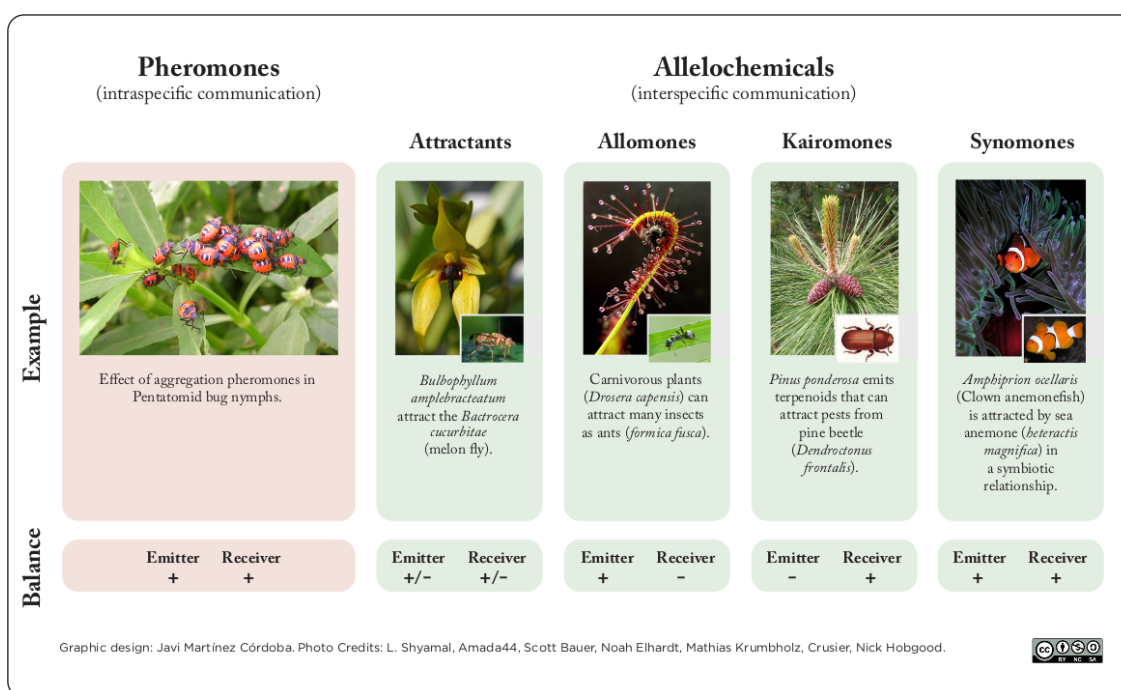


Figure 1: **Infochemical communication is classified by its adaptive function.** Pheromones mediate in intraspecific communication, being beneficial to individuals of the species, while allelochemicals intercede in interspecific communication. Allelochemicals are classified into four subcategories: allomones, which benefit the emitter but not the receiver; kairomones, profiting the receiver but not the emitter; synomones, which satisfy both sender and receiver in symbiotic interactions; and attractants which, depending on the case, can benefit or harm either the sender or the receiver.

39 Historically, researchers have established parallels with Linguistics intending to improve the under-  
 40 standing of chemical communication. One paradigmatic case is the description of DNA structure  
 41 [14, 15] where similarities between the genetic code and the verbal language have been validated  
 42 [16]. Language patterns can be theorized and quantitatively measured with the use of linguistic  
 43 laws, statistical regularities shared across human communication [17, 18], and also discovered in  
 44 biological systems [19] as diverse as the vocalizations [20, 21, 22, 23, 24] and gestures of other pri-  
 45 mates [25], genomics [26, 27], proteomics [28, 29, 30, 31] and chemical communication systems [32].

46

47 One important linguistic law is Zipf's law of brevity [33, 34], or just Brevity law: the tendency for  
 48 the most frequent communicative elements to be shorter or briefer [18]. This law is a consequence  
 49 of the so-called principle of compression [35] which in standard information theory consists of  
 50 assigning codes as short as possible to the more frequently occurring symbols and longer codewords  
 51 to the symbols that infrequently occur [36]. In this theoretical framework, optimal non-singular

52 coding predicts that in language, the length of a word should grow approximately as the logarithm  
 53 of its frequency of occurrence [36]. Thus, if  $f$  is the frequency of a communicative element and  $\ell$   
 54 is its length, Zipf's law of brevity can be formulated as [18]:

$$f \sim \exp(-\lambda\ell), \quad \lambda > 0 \quad (1)$$

55 where  $\lambda$  is a fitting parameter.

56  
 57 In chemical communication, Zipf's law of brevity predicts that the more frequent a chemical com-  
 58 pound is, the lighter and shorter. The size of infochemicals can be measured by considering their  
 59 molecular weight (MW) or by counting their number of carbons (NC), which, in general, correlate  
 60 together [37]. Therefore, it would be expected that the most frequent infochemicals have low  
 61 MW and are composed of fewer carbons. Many species share the same chemical substances in  
 62 their communication systems [4, 38, 39], so we can define the degree of use [73] of an infochemi-  
 63 cal as the number of species that interacts with it. Therefore, from an ecological perspective, the  
 64 infochemical degree of use would be equivalent to the total number of species that are sensitive to it.

65  
 66 For the first time, we empirically explore and discuss whether Zipf's law of brevity and compression  
 67 principle hold in the infochemical communication system that governs much of the communication  
 68 between insects, plants, and other organisms. To address this critical issue, here we conduct a  
 69 systematic exploration of Pherobase, the largest available infochemical database [39]. We will  
 70 show that Zipf's law of brevity is fulfilled for allelochemicals –interspecific communication– but  
 71 not for pheromones –intraspecific communication–, suggesting that chemical communication be-  
 72 tween individuals of the same species is strongly influenced by other factors beyond the principle  
 73 of compression.

## 75 2 Materials and methods

76 Information was fetched from Pherobase, the largest and most comprehensive open-access info-  
 77 chemical database [39]. Pherobase contains elaborate and verified infochemical information, mak-  
 78 ing it a very reliable and valuable resource widely used by the scientific community [40, 41, 42].  
 79 With 36,133 recorded interactions for 1,393 different infochemicals at the time of this study [72]  
 80 – see Table 1 –, each infochemical is classified with details on what its adaptive function is, de-  
 81 pending on the species interacting with it [39]. Among them, the statistical preponderance of  
 82 insects (34.55%) and arachnids (7.61%) is remarkable (see section 1 of the Supplementary Infor-  
 83 mation for more details on the most represented taxonomies). The same infochemical (type) can  
 84 be used by several species in the same or different ecosystem with different functions. Therefore,  
 85 the aggregation of sub-types (e.g. Attractants, Allomones) is not necessarily equal to the total  
 86 number of types of the aggregation (e.g. Allelochemicals) –see  $r$  in the third column of Table 1–.  
 87 For all infochemicals, it is obtained the molecular weight (MW), chemical composition, number  
 88 of carbons (NC), and the number of species (degree of use) that interact with the compound, in  
 89 each specific adaptive function – see Figure 1 –. The group of Synonomes was excluded from the  
 90 research since only three different items were reported.

91  
 92 Firstly, it is explored the probability distribution of infochemical sizes considering MW and NC  
 93 as study units. As it will be shown the probability distributions are subexponential heavy-tailed  
 94 [43], so lognormal (LND)  $p(x, \mu, \sigma) = \frac{1}{x\sigma\sqrt{2\pi}} e^{-\frac{(\ln x - \mu)^2}{2\sigma^2}}$ , gamma  $p(x, \kappa, \theta) = \frac{1}{\Gamma(\kappa)\theta^\kappa} x^{\kappa-1} e^{-\frac{x}{\theta}}$  – and  
 95 Weibull  $p(x, k, \lambda) = \frac{k}{\lambda} \left(\frac{x}{\lambda}\right)^{k-1} e^{-(x/\lambda)^k}$  candidate distributions are fitted to data by maximum like-  
 96 lihood estimation method [44]. Kolmogorov-Smirnov test is carried out for goodness of fit and  
 97 mean loglikelihood is used for model selection. Then, following [45], one-sided Kendall corre-  
 98 lation test [46] is used to assess if there is a negative monotonic relationship between the use  
 99 degree of infochemicals and the compound size, considering MW and NC. Additionally, Spearman

<http://mc.manuscriptcentral.com/bl>

	$N$	$r$	$H(r)$	degree-MW		degree-NC	
				$\tau$	$p$ -value	$\tau$	$p$ -value
All infochemicals	36 133	1 393	9.0	-0.11	$p < 10^{-4}$	-0.09	$p < 10^{-4}$
Pheromones	18 973	1 178	8.9	-0.01	$p = 0.36$	0.01	$p = 0.74$
Alelochemicals	17 063	829	8.1	-0.22	$p < 10^{-4}$	-0.19	$p < 10^{-4}$
Attractants	10 761	531	7.4	-0.14	$p < 10^{-4}$	-0.13	$p < 10^{-4}$
Allomones	5 654	500	7.3	-0.21	$p < 10^{-4}$	-0.21	$p < 10^{-4}$
Kairomones	642	201	7	-0.22	$p < 10^{-4}$	-0.15	$p < 10^{-2}$

Table 1: **Parameters and correlations across infochemical adaptive function.** Total number of interactions ( $N$ ), number of different infochemicals ( $r$ ), entropy of the rank empirical distribution function ( $H(r)$ ) and one sided Kendall correlation ( $\tau$ ) between degree of use and infochemical size both both in continuous magnitudes –Molecular Weight ( $MW$ )– and in discrete units –Number of Carbons ( $NC$ )–.

100 and Pearson correlation tests are also provided in section 2 of the Supplementary Information (SI).

101

102 We also explore the level of information contained on the probability distributions of ranks by an-  
 103 analyzing the Shannon entropy  $H(r) = -\sum_{i=1}^n P(r_i) \log_2 P(r_i)$  [47], where  $P(r_i)$  is the probability  
 104 degree associated of the infochemical with rank  $r$ , after sorting infochemical type by use degree in  
 105 descending order and being  $r = 1$  the most frequent one [33, 34]. Alternative entropy estimators  
 106 are provided and discussed in section 7 of SI [48]. Finally, the mean square error method is used  
 107 to fit equation 1 to assess the relationship between infochemical size and degree.

108

## 109 3 Results

### 110 3.1 Degree and allelochemical size are negatively correlated

111 We start by analysing the Pherobase dataset to address if there is a negative correlation between  
 112 the size of infochemicals and their degree of use. The size is measured using two different mag-  
 113 nitudes: MW and NC. The first can be approximated as a continuous magnitude, while the last  
 114 can be understood as a discrete unit. Given that choosing the magnitude of the study is still an  
 115 open problem and that both alternatives are related to the energy used in communication, this  
 116 approach will allow us to address recent discussions in the study of linguistic laws [18, 49, 50].

117

118 For a given adaptive function –e.g. Pheromones– and following [45], we compute a one-sided  
 119 Kendall correlation test ( $\tau$ ) to explore manifestations of the compression principle (Spearman and  
 120 Pearson are also considered in section 2 of SI). Our results are summarised in Table 1, revealing  
 121 a strong negative correlation between infochemical size and degree of use when all infochemi-  
 122 cals are studied together. Different behaviours are observed when the subtypes of infochemi-  
 123 cals are assessed. Interestingly, Zipf’s law of brevity is fulfilled in interspecies communication  
 124 –allelochemicals–, while not in Pheromones. Moreover, allelochemicals are also studied with more  
 125 granularity, finding likewise strong negative correlation regardless of their adaptive function. These  
 126 results are robust for both units of study: MW and NC. Additionally, the Shannon entropy of rank  
 127 probability distribution has been estimated for each adaptive group, uncovering that pheromones  
 128 are more entropic than allelochemicals subgroups.

129

### 130 3.2 Infochemical size is lognormally distributed when using MW

131 Following previous studies in human communication [18, 49], we explore the marginal distribution  
 132 of infochemical size considering MW and NC as units of study. To begin with, using the method  
 133 of maximum-likelihood estimation (MLE), we have fitted the data to three possible theoretical

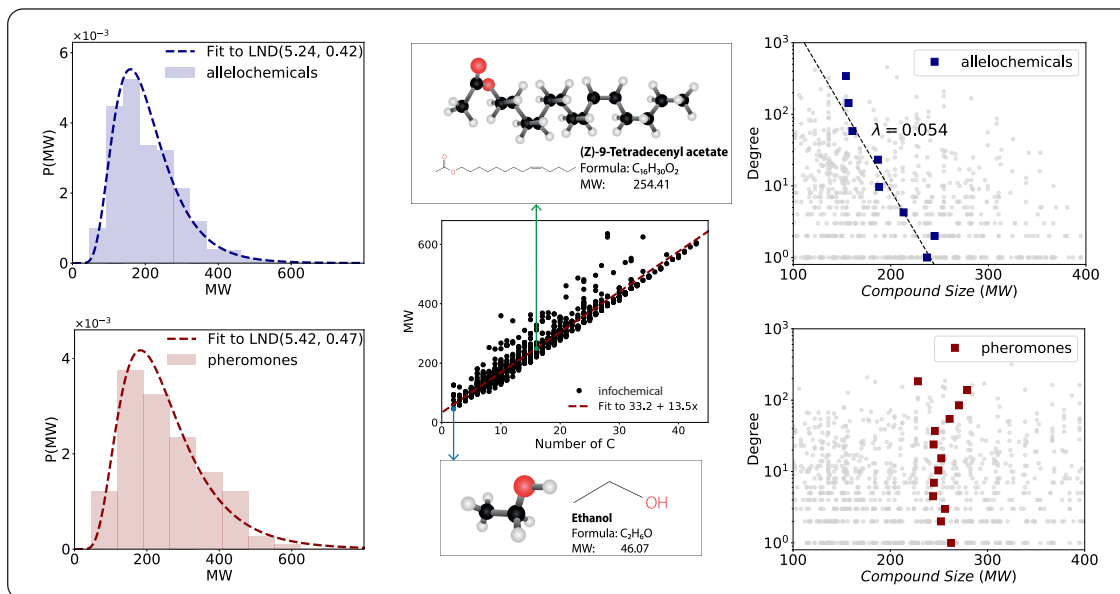


Figure 2: **Brevity law in infochemical communication.** First two left panels show the molecular weight distribution of allelochemicals (blue bars) and pheromones (red bars) with fitting to best LND distribution (dotted line). The central panel illustrates the almost linear relationship between the Number of Carbons (NC) and Molecular Weight (MW), with two illustrative examples of infochemicals, including the most frequent one: Ethanol. Finally, the right panels reveal that Brevity law holds for interspecies communication –allelochemicals– while not for intraspecific communication –pheromones–.

134 distributions: LND, gamma and Weibull (we use Kolmogorov–Smirnov distance  $ks$  for goodness  
 135 of fits, and mean loglikelihood  $\langle \mathcal{L} \rangle$  for model selection, see table 2 and, additionally, tables 4 and  
 136 5 of SI). Jointly considering  $\langle \mathcal{L} \rangle$  and  $ks$ , MW distribution is best explained by LNDs, in line with  
 137 results reported in [18] where continuous physical magnitudes were used. Left panels of Figure 2  
 138 show estimated MW distribution for allelochemicals (blue) and for pheromones (red). Colour bars  
 139 correspond to empirical binned data, and dotted lines are the MLE fit to LND. Nonetheless, the  
 140 gamma probability distribution is the most plausible when considering NC –discrete magnitudes–  
 141 for addressing the probability size distribution of infochemicals. These results agree with previ-  
 142 ously reported investigations where the authors examined word length distribution in texts with  
 143 discrete units of study [49].

144

### 145 3.3 MW and NC are strongly correlated

146 The appropriate choice of study units to explore compression principle and Brevity law remains  
 147 an open question that should be carefully considered. Nevertheless, the chosen magnitudes should  
 148 contain information on the energy and communication effort intended to be minimized. Here, we  
 149 have considered both a discrete unit of study –NC– and a continuous magnitude –MW–, and we  
 150 explore their relationship. In the central panel of Figure 2 each infochemical is represented with  
 151 a black circle. Dotted red line represents the linear regression between the MW and NC of each  
 152 infochemical. It has been found that MW and NC are strongly correlated, so both magnitudes  
 153 should be appropriate for exploring manifestations of the compression principle.

154

	Type	LND				gamma				Weibull	
		$\langle \mathcal{L} \rangle$	$D_{ks}$	$\mu$	$\sigma$	$\langle \mathcal{L} \rangle$	$D_{ks}$	$\kappa$	$\theta$	$\langle \mathcal{L} \rangle$	$D_{ks}$
MW	All infochemicals	-6.05	0.05	5.4	0.46	-6.06	0.06	4.89	0.02	-6.10	0.06
	Pheromones	-6.09	0.05	5.43	0.47	-6.09	0.06	4.84	0.019	-6.13	0.05
	Allelochemicals	-5.78	0.06	5.24	0.42	-5.79	0.08	6.01	0.029	-5.84	0.08
	Attractans	-5.69	0.07	5.20	0.39	-5.69	0.09	6.74	0.034	-5.73	0.1
	Allomones	-5.78	0.08	5.22	0.42	-5.80	0.09	5.67	0.028	-5.86	0.11
	Kairomones	-5.61	0.15	5.03	0.43	-5.66	0.18	5.22	0.031	-5.77	0.19
NC	All infochemicals	-3.44	0.06	2.61	0.55	-3.42	0.08	3.71	0.23	-3.44	0.08
	Pheromones	-3.49	0.07	2.66	0.56	-3.47	0.08	3.71	0.22	-3.48	0.08
	Allelochemicals	-3.14	0.08	2.43	0.49	-3.13	0.09	4.54	0.35	-3.16	0.1
	Attractans	-3.11	0.1	2.41	0.48	-3.08	0.1	4.83	0.39	-3.09	0.12
	Allomones	-3.11	0.09	2.39	0.49	-3.11	0.11	4.41	0.35	-3.15	0.12
	Kairomones	-3.00	0.16	2.23	0.52	-3.01	0.19	3.92	0.37	-3.06	0.21

Table 2: **Estimated parameters of infochemical size distribution.** Probability distribution candidates are fitted using MLE, model selection is based on maximizing mean log-likelihood  $\langle \mathcal{L} \rangle$  and goodness of fit is based on Kolmogorov-Smirnov distance  $D_{ks}$ , where lower values mean that data are more plausible to follow the distribution. The estimated parameters for LND( $\mu, \sigma$ ) and gamma( $\kappa, \theta$ ) are provided.

### 3.4 Infochemical degree of use follows Brevity law –except for pheromones–

To round off, we finally consider Zipf’s law of brevity in detail. In the right panels of Figure 2, we scatter plot MW versus degree of use of infochemicals with grey circular dots. Then, we apply a logarithmic binning over the degree axis (blue for allelochemicals and red for pheromones) in order to counterbalance low sampling effects [18]. Other infochemical subgroups are also considered in Figure 3 of SI. Data is then fitted to equation 1 and represented with a red dotted line showing an excellent agreement with the square colour bins used as visual information. Pheromones are not fitted to equation 1 because in section 3.1, it was previously shown that no correlation was held.

## 4 Discussion

We have validated with unprecedented detail, by studying more than 36,000 infochemical interactions, that Zipf’s law of brevity operates at interspecies infochemical communication. Those results are in line with precedent studies in language, and other communication systems [18, 20, 21, 22, 51, 52, 53], suggesting the universality of this linguistic law and the principle of compression, from which it derives. Nevertheless, we have found no evidence that Zipf’s law of brevity holds in intraspecific chemical communication –pheromones–.

Previous work pointed out that communication efficiency is equivalent to minimising the length  $l$  of a communicative element [27, 35, 54]. In contrast, unambiguity and maximum accuracy in communication is analogous to a non-singular scheme of communication, where a unique element is assigned to each meaning [36, 45]. Pheromones may be statistically uniformly distributed so that even if there were optimal coding [36, 45], it would not manifest itself in the form of Zipf’s law of brevity. The failure to obtain a significant p-value for the pheromones (see Table 1) does not mean that Zipf’s law of brevity is not operating but raises the fact that other communicative principles are present [35, 55].

In some noisy or adverse domains, selection pressures have maximized transmission through redundancy in opposite to efficiency or energy saving, as it happens in acoustic long-distance calls [50, 55, 56]. This could be happening in the case of intraspecific pheromone communication,

185 and it is consistent with the unrevealed Shannon entropy, significantly higher when comparing  
186 pheromones to other allelochemicals [36, 57].

187  
188 The Handicap principle goes even further regarding the energy effort spent on communication  
189 [58]. It suggests that some individuals may expend even more effort than necessary as long as  
190 they demonstrate that others can not afford it. The so-called handicap hypothesis (see [59] about  
191 terminology) has evolved considerably from the original works [60, 61], early criticized [62, 63],  
192 and discussed for its lack of evidence [59]. The fact that pheromones do not comply with Zipf's  
193 law of brevity and their higher entropy may be an indirect evidence of the Handicap principle.

194  
195 Pheromones have an intraspecific communication function, including a sexual role, but these uses  
196 are not granular detailed in the Pherobase dataset. Nevertheless, concerning sex pheromones,  
197 they can attract both the attention of predators and potential mates [60, 65, 66]. Natural selec-  
198 tion may favour the evolution of chemical compounds whose reproductive benefits outweigh the  
199 adverse effects on survival so that the extra cost of increasing the chemical length of a substance  
200 may be offset by making it harder to be detected by predators [59, 66, 65]. For this reason, the  
201 size increase of pheromones could be a consequence of evolutionary survival pressures. However,  
202 the factors affecting signalling during communication are manifold and can significantly skew the  
203 direction of evolutionary pressures (see [66] for a review).

204  
205 On the other hand, considering that pheromones are issued in parallel to other communication  
206 channels [4], the communicative complexity [38] and the redundancy of pheromone signals [64] may  
207 also explain the non-compliance with Zipf's law of brevity. Honeybee queens emit multiple redun-  
208 dant signals to ensure the message is correctly perceived, avoiding possible signal detection and  
209 identification errors [64]. In contrast, the parallel study of ubiquitous substances in many species  
210 and ecosystems may be essential to understanding interkingdom communication (e.g. melatonin  
211 case in [67]).

212  
213 Moreover, at the outset, we have explored the influence of the units of study to measure the  
214 manifestation of Zipf's law of brevity in chemical communication. We have obtained a LND  
215 when analysing the infochemicals size distribution when measuring it with MW –a continuous  
216 magnitude– but when considering the NC –a discrete unit–, the gamma distribution is more plau-  
217 sible (Table 2). Previous works suggest that the functional form of Zipf's law of brevity depends  
218 crucially on the length distribution function of the communicative elements, and they obtained  
219 a gamma distribution for text characters –a discrete magnitude– [49]. On the other hand, in a  
220 previous study of Zipf's law of brevity in speech, considering for length a continuous unit of study,  
221 LND was more plausible [18]. Our findings are consistent with both results, and further work  
222 is needed to elucidate the influence of study units on signal size distributions in communication  
223 systems.

224  
225 Interestingly, it has been shown that both magnitudes of study are strongly correlated, finding a  
226 linear correlation between the NC and MW. This is explained by the tendency of infochemicals to  
227 be carbon-chain compounds [32, 37]. This result would suggest both magnitudes are equivalent,  
228 but in general, it is theoretically preferable to deal with continuous physical quantities as close as  
229 possible to the observable reality [18].

230  
231 Finally, the conclusions reached by this work are shaped by the limitations of the Pherobase  
232 dataset itself. This dataset is based on scientific evidence, so many infochemicals are still not  
233 reported; therefore, the degree of use of specific infochemicals may be underestimated. Besides,  
234 some taxonomic groups prevail in Pherobase, as reported in Section 1 of the SI. This fact, how-  
235 ever, does not detract from the validity of the overall study carried out here, although a detailed  
236 analysis of the different families, classes and orders should be further addressed in the future.

237  
238 Future investigations should consider a detailed, granular exploration of infochemical communi-

239 cation within ecosystems to determine the dynamic of ecological interactions that it is happening  
240 in each case. Although this study has considered infochemical communication from a global  
241 perspective, beyond being a limitation, it could be instrumental in exobiology. Exobiology is a  
242 discipline where very diverse approaches do necessarily coexist in the search for life outside the  
243 Earth [68, 69]. We suggest that the infochemicals distribution on Earth can be a comparative  
244 indicator to detect life on other planets, and the presence of specific carbon chains an indicative  
245 of chemical communication [70, 71].

## 246 Data accessibility

247 Pherobase is a freely accessible database. Curated data used in this study containing info-  
248 chemical information is now freely available in Dryad Digital Repository <https://doi.org/10.5061/dryad.905qfttnx> [72]. All the necessary scripts and coding used during this research for  
249 processing data, computing and figure generation are freely available in <https://github.com/ivangtorre/compression-and-brevity-law-in-infochemical-communication>. We have used  
250 Python 3.8 and R 4.1 with fitdistrplus library for the analysis. Other libraries such as Numpy  
251 1.22, Pandas 1.0 or Matplotlib 3.2 are also used.

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