A cross-language perspective on speech information rate

François Pellegrino, Christophe Coupé and Egidio Marsico

Laboratoire Dynamique Du Langage, Université de Lyon,
Centre National de la Recherche Scientifique

François Pellegrino
DDL - ISH
14 Avenue Berthelot
69363 Lyon Cedex 7
France.
francois.pellegrino@univ-lyon2.fr
+33-4-72-72-64-94
A cross-language perspective on speech information rate

Abstract

This paper cross-linguistically investigates the hypothesis that the average information rate conveyed during speech communication results from a trade-off between average information density and speech rate. The study, based on 7 languages, shows a negative correlation between density and rate illustrating the existence of several encoding strategies. However these strategies do not necessarily lead to a constant information rate. These results are further investigated in relation with the notion of syllabic complexity1.

Keywords: speech communication, information theory, working memory, speech rate, cross-language study, British English, French, German, Italian, Japanese, Mandarin Chinese, Spanish

1 We wish to thank C.-P. Au, S. Blandin, E. Castelli, S. Makioka, G. Peng and K. Tamaoka for their help with collecting or sharing the data. We also thank Fermín Moscoso del Prado Martin, R. Harald Baayen, Barbara Davis, Peter Mac Neilage, Michael Studdert-Kennedy, and two anonymous reviewers for their constructive criticism and their suggestions on earlier versions of this paper.
1. INTRODUCTION.

“As soon as human beings began to make systematic observations about one another's languages, they were probably impressed by the paradox that all languages are in some fundamental sense one and the same, and yet they are also strikingly different from one another.” (Charles A. Ferguson, 1978).

Ferguson's quotation describes two goals of linguistic typology: searching for invariants and determining the range of variation found across languages. Invariants are supposed to be a set of compulsory characteristics, which presumably defines the core properties of the language capacity itself. Language being a system, invariants can be considered as systemic constraints imposing a set of possible structures among which languages ‘choose’. Variants can then be seen as language strategies compatible with the degrees of freedom in the linguistic constraints. Both directions are generally investigated simultaneously as the search for universals contrastively reveals the differences. Yet, linguistic typology has mostly revealed that languages vary to a large extent, finding only few, if any, absolute universals, unable to explain how all languages are "one and the same" and reinforcing the fact that they are "so strikingly different"(see Evans and Levinson (2009) for a recent discussion). Nevertheless, the paradox only exists if one considers both assumptions ("one and the same" and "strikingly different") at the same level. Language is actually a communicative system which primary function is to transmit information. The unity of all languages is probably to be found in this function, regardless of the different linguistic strategies on which they rely on.

Another well-known assumption is that all human languages are overall equally complex. This statement is present in most introductory classes in linguistics or encyclopaedias (e.g., see
Crystal (1987)). At the same time, linguistic typology provides extensive evidence that the complexity of each component of language grammar (phonology, morphology or syntax) widely varies from one language to another, and nobody claims that two languages with 11 vs. 141 phonemes (like respectively Rotokas and !Xu) are of equal complexity with respect to their phonological systems (Maddieson, 1984). A balance in complexity must therefore operate within the grammar of each language: a language exhibiting a low complexity in some of its components should compensate with a high complexity in others. As exciting as this assumption looks, no definitive argument has yet been provided to support or invalidate it (see discussion in Planck (1998)), even if a wide range of scattered indices of complexity have recently come into sight, and so far led to partial results in a typological perspective (Cysouw (2005); Dahl (2004); Fenck-Oczlon and Fenck (1999, 2005); Maddieson (2006, 2009); Marsico et al. (2004); Shosted (2006)) or from an evolutionary viewpoint (see Sampson, Gil & Trudgill, 2009, for a recent discussion).

Considering that the communicative role of language has been underestimated in those debates, we suggest that the assumption of an “equal overall complexity” is ill-defined. More precisely, we endorse that all languages exhibit an “equal overall communicative capacity” even if they have developed distinct encoding strategies partly illustrated by distinct complexities in their linguistic description. This communicative capacity is probably delimited within a range of possible variation in terms of rate of information transmission: below a lower limit, speech communication would not be efficient enough to be socially useful and acceptable; above an upper limit, it would exceed the human physiological and cognitive capacities. One can thus postulate an optimal balance between social and cognitive constraints, taking also the characteristics of transmission along the audio channel into account.
This hypothesis predicts that languages are able to convey relevant pragmatic-semantic information at similar rates and urges to pay attention to the rate of information transmitted during speech communication. Studying the encoding strategy (as revealed by an information-based and complexity-based study, see below) is thus one necessary part of the equation but it is not sufficient to determine the actual rate of information transmitted during speech communication.

After giving some historical landmarks on the way the notions of information and complexity have been interrelated in linguistics for almost one century (Section 2), this article aims at putting together the information-based approach and the cross-language investigation. It cross-linguistically investigates the hypothesis that a trade-off is operating between a syllable-based average information density and the rate of transmission of syllables in human communication (Section 3). The study, based on comparable speech data from 7 languages, provides strong arguments in favour of this hypothesis. The corollary assumption predicting a constant average information rate among languages is also examined. An additional investigation of the interactions between these information-based indices and a syllable-based measure of phonological complexity is then provided to extend the discussion toward future directions, in the light of literature on least-effort principle and cognitive processing (Section 4).

2. HISTORICAL BACKGROUND.

The concept of information and the question of its embodiment in linguistic forms were implicitly introduced in linguistics at the beginning of the 20th century, even before the so-called Information Theory was popularized (Shannon and Weaver 1949). They were first addressed to the light of approaches such as the frequency of use (from Zipf (1935), to Bell et al. (2009)) or the functional load (from Martinet (1933), and Twaddell (1935), to Surendran and Levow
Starting from the 1950’s, they then benefited from inputs from Information Theory, with notions such as entropy, communication channel and redundancy\(^2\) (Cherry et al. (1953), Hockett (1953), Jakobson and Halle (1956), inter alia).

Furthermore, in the quest for explanations of linguistic patterns and structures, the relationship between information and complexity has also been addressed, either synchronically or diachronically. A landmark is given by Zipf stating that ‘(...) there exists an equilibrium between the magnitude or degree of complexity of a phoneme and the relative frequency of its occurrence’ (Zipf 1935: 49). Trubetzkoy and Joos strongly attacked this assumption: in the Grundzüge, Trubetzkoy denied any explanatory power to the uncertain notion of complexity in favor of the notion of markedness (Trubetzkoy 1938) while Joos’s criticism focused mainly on methodological shortcomings and what he considered a tautological analysis (Joos (1936); but see also Zipf’s answer (1937)).

Later, the potential role of complexity in shaping languages has been discussed either by its identification with markedness or by considering it in a more functional framework. The first tendency is illustrated by Greenberg answering to the self-question ‘Are there any properties which distinguish favored articulations as a group from their alternatives?’ by ‘the principle that of two sounds that one is favored which is the less complex’. He then concluded that ‘the more complex, less favored alternative is called marked and the less complex, more favored alternative the unmarked’ (Greenberg 1969: 476–477). The second approach, initiated by Zipf’s principle of least-effort, has been developed by considering that complexity and information may play a role in the regulation of linguistic systems and speech communication. While Zipf mostly ignored the listener’s side and suggested that the least-effort was almost exclusively a constraint affecting the speaker, more recent works demonstrated that other forces also play an important role and that
economy or equilibrium principles result from a more complex pattern of conflicting pressures (e.g. Martinet (1955, 1962); Lindblom (1990)). For instance, Martinet emphasized the role of the communicative need (‘the need for the speaker to convey his message’ (Martinet, 1962:139)), counterbalancing the principle of speaker’s least effort. Lindblom’s H&H theory integrates a similar postulate, leading to self-organizing approaches to language evolution (e.g. Oudeyer (2006)) and to taking the listener’s effort into consideration.

More recently, several theoretical models have been proposed to account for this regularity and to reanalyse Zipf’s assumption in terms of emergent properties (e.g. Ferrer i Cancho, 2005; Ferrer i Cancho and Solé, 2003; Kuperman et al. 2008). These recent works strongly contribute to a renewal of information-based approaches to human communication (along with Aylett and Turk (2004); Frank and Jaeger (2008); Genzel and Charniak (2003); Goldsmith (2000, 2002); Harris (2005); Hume (2006); Keller (2004); Maddieson (2006); Pellegrino et al. (2007); van Son and Pols (2003), inter alia), but mostly in language-specific studies (see however Kuperman et al., 2008 and Plantadosi, Tily, and Gibson, 2009).

3. SPEECH INFORMATION RATE

3.1. MATERIAL. The goal of this study is to assess whether there exist differences in the rate of information transmitted during speech communication in several languages. The proposed procedure is based on a cross-language comparison of the speech rate and the information density of seven languages using comparable speech materials. Speech data are a subset of the MULTEXT multilingual corpus (Campione & Véronis (1998); Komatsu et al. (2004)). This subset consists of $K = 20$ texts composed in British English, freely translated into the following languages to convey a comparable semantic content: French (FR), German (GE), Italian (IT),
Japanese (JA), Mandarin Chinese (MA), and Spanish (SP). Each text is made of five semantically connected sentences composing either a narration or a query (to order food by phone, for example). The translation inevitably introduced some variation from one language to another, mostly in named entities (locations, etc.) and to some extent in lexical items, in order to avoid odd and unnatural sentences. For each language, a native or highly proficient speaker counted the number of syllables in each text, as uttered in careful speech, as well as the number of words, according to language-specific rules. The Appendix gives the version of one of the 20 texts in the seven languages, as an example.

Several adult speakers (from six to ten, depending on the language) recorded the 20 texts at “normal” speech rates, without being asked to produce fast or careful speech. No sociolinguistic information on them is provided with the distributed corpus. 59 speakers (29 male and 30 female speakers) of the seven target languages were included in this study, for a total number of 585 recordings and an overall duration of about 150 minutes. The text durations were computed discarding silence intervals longer than 150 ms, according to a manual labelling of speech activity.3

Since the texts were not explicitly designed for detailed cross-language comparison, they exhibit a rather large variation in length. For instance, the lengths of the 20 English texts range from 62 to 104 syllables. To deal with this variation, each text was matched with its translation in an eighth language, Vietnamese (VI), different from the seven languages of the corpus. This external point of reference was used to normalize the parameters for each text in each language and consequently to facilitate the interpretation by comparison with a mostly isolating language (see below).
The fact that this corpus was composed of read-aloud texts, which is not typical of natural speech communication, can be seen as a weakness. Though the texts mimicked different styles (ranging from very formal oral reports to more informal phone queries), this procedure most likely underestimated the natural variation encountered in social interactions. Reading probably lessens the impact of paralinguistic parameters such as attitudes and emotions and smooths over their prosodic correlates (e.g. Johns-Lewis, 1986). Another major and obvious change induced by this procedure is that the speaker has no leeway to choose his/her own words to communicate, with the consequence that a major source of individual, psychological and social information is absent (Pennebaker, Mehl and Niederhoffer, 2003). Recording bilinguals may provide a direction for future research on cross-linguistic differences in speech rates while controlling for individual variation. However, this drawback may also be seen as an advantage since all the 59 speakers of the 7 languages are recorded in similar experimental conditions, leading to comparable data.

3.2. DENSITY OF SEMANTIC INFORMATION. In the present study, density of information refers to the way languages encode semantic information in the speech signal. In this view, a dense language will make use of fewer speech chunks than a sparser language for a given amount of semantic information. This section introduces a methodology to evaluate this density and to further assess whether information rate varies from one language to another.

Language grammars reflect conventionalized language-specific strategies for encoding semantic information. These strategies encompass more or less complex surface structures and more or less semantically transparent mappings from meanings to forms (leading to potential trade-offs in terms of complexity or efficiency, see for instance Dahl (2004) and Hawkins (2004, 2009)), and they output meaningful sequences of words. The word level is at the heart of human communication, at least because of its obvious function in speaker–listener interactions and also
because of its central status between meaning and signal. Thus, words are widely regarded as the relevant level to disentangle the forces involved in complexity trade-offs and to study the linguistic coding of information. For instance, Juola applied information-theoretical metrics to quantify the cross-linguistic differences and the balance between morphology and syntax in the meaning-to-form mapping (Juola, 1998, 2008). At a different level, van Son and Pols, among others, have investigated the form-to-signal mapping, viz. the impact of the linguistic information distribution on the realized sequence of phones (van Son and Pols, 2003, 2005; see also Aylett and Turk, 2006). These two broad issues (from meaning to form, and from form to signal) shed light on the constraints, the degrees of freedom, and the trade-offs that shape human languages. In this study, we propose a different approach that focuses on the direct mapping from meaning to signal. More precisely, we focus on the level of the information encoded in the course of the speech flow. We hypothesize that a balance between the information carried by speech units and their rate of transmission may be observed, whatever the linguistic strategy of mapping from meaning to words (or forms) and from words to signals.

Our methodology is consequently based on evaluating the average density of information in speech chunks. The relationship between this hypothetical trade-off at the signal level and the interactions at play at the meaningful word level is an exciting topic for further investigation; it is however beyond the scope of this study.

The first step is to determine the chunk to use as a ‘unit of speech’ for the computation of the average information density per unit in each language. Units such as features or articulatory gestures are involved in complex multidimensional patterns (gestural scores or feature matrices) not appropriate for computing the average information density in the course of speech communication. On the contrary, each speech sample can be described in terms of discrete
sequences of segments or syllables; these units are possible candidates, though their exact status and role in communication is still questionable (e.g., see Port and Leary (2005) for a criticism of the discrete nature of those units). This study is thus based on syllables for both methodological and theoretical motivations (see also Section 3.3).

Assuming that for each text $T_k$, composed of $\sigma_k(L)$ syllables in language $L$ the overall semantic content $S_k$ is equivalent from one language to another, the average quantity of information per syllable for $T_k$ and for language $L$ is:

$$I^k_L = \frac{S_k}{\sigma_k(L)}$$

Since $S_k$ is language-independent, it was eliminated by computing a normalized Information Density $ID$ using VI as the benchmark. For each text $T_k$ and language $L$, $ID^k_L$ resulted from a pairwise comparison of the text lengths (in terms of syllables) respectively in $L$ and VI:

$$ID^k_L = \frac{I^k_L}{I^k_{VI}} = \frac{S_k}{\sigma_k(L)} \times \frac{\sigma_k(VI)}{S_k} = \frac{\sigma_k(VI)}{\sigma_k(L)}$$

Next, the average information density $ID_L$ (in terms of linguistic information per syllable) with reference to VI is defined as the mean of $ID^k_L$ evaluated for the $K$ texts:

$$ID_L = \frac{1}{K} \sum_{k=1}^{K} ID^k_L$$

If $ID_L$ is superior to unity, $L$ is “denser” than VI since on average fewer syllables are required to convey the same semantic content. An $ID_L$ lower than unity indicates, on the contrary, that $L$ is not as dense as VI.
The averaging over 20 texts aimed at getting values pointing towards language-specific grammars rather than artefacts due to idiomatic or lexical biases in the constitution of the texts. On average among the 8 languages, each text consists of 102 syllables, for a total number of syllables per language of 2,040, which is a reasonable length to estimate central tendencies such as means or medians. Another strategy, used by Fenk-Oczlon & Fenk (1999) is to develop a comparative database made of a set of short and simple declarative sentences (22 in their study) translated in each of the language considered. Their option was that using simple syntactic structure and very common vocabulary results in a kind of baseline suitable to proceed to the cross-language comparison without bias such as stylistic variation. However, such short sentences (ranging on average in Fenk-Oczlon and Fenk database from 5 to 10 syllables per sentence, depending on the language) could be more sensitive to lexical bias than longer texts, resulting in wider confidence intervals in the estimation of information density.

Table 1 (second column) gives the $ID_L$ values for each of the seven languages. The fact that Mandarin exhibits the closest value to Vietnamese ($ID_{MA} = 0.94 \pm 0.04$) is compatible with their proximity in terms of lexicon, morphology and syntax. Furthermore, Vietnamese and Mandarin, which are the two tone languages of this sample, reach the highest values. According to our definition of density, Japanese density is one-half of the Vietnamese reference ($ID_{JA} = 0.49 \pm 0.02$). Consequently, even in this small sample of languages, $ID_L$ exhibits a considerable range of variation, reflecting different grammars.

These grammars reflect language-specific strategies for encoding linguistic information but they ignore the temporal facet of communication. For example, if the syllabic speech rate (i.e. the average number of syllables uttered by second) is twice as fast in Japanese as in Vietnamese, the linguistic information would be transmitted at the same rate in the two...
languages, since their respective Information densities per syllable $ID_{JA}$ and $ID_{VI}$ are inversely related. In this perspective, linguistic encoding is only one part of the equation and we propose in the next section to take the temporal dimension into account.

**INSERT TABLE 1 HERE**

3.3. **VARIATION IN SPEECH RATE.** Roach (1999) claimed that the existence of cross-language variations of speech rate is one of the language myths, due to artefacts in the communication environment or its parameters. However, he considered that syllabic rate is a matter of syllable structure and is consequently widely variable from one language to another, leading to perceptual differences: ‘So if a language with a relatively simple syllable structure like Japanese is able to fit more syllables into a second than a language with a complex syllable structure such as English or Polish, it will probably sound faster as a result’ (Roach 1999). Consequently, Roach proposed to estimate speech rate in terms of sounds per second, to depart from this subjective dimension. However, he immediately identified additional difficulties in terms of sound counting, due for instance to adaptation observed in fast speech: ‘The faster we speak, the more sounds we leave out’ (Roach 1999). On the contrary, the syllable is well known for its relative robustness during speech communication: Greenberg (1999) reported that syllable omission was observed for about 1% of the syllables in the Switchboard corpus while omissions occur for 22% of the segments. Using a subset of the Buckeye corpus of conversational speech (Pitt et al., 2005), Johnson (2004) found a higher proportion of syllable omissions (5.1% on average) and a similar proportion of segment omissions (20%). The difference observed in terms of syllable deletion rate may be due to the different recording conditions: Switchboard data
consist of short conversations on the telephone while the Buckeye corpus is based on face-to-
face interaction during longer interviews. The latter is more conducive to reduction for at least
two reasons: multimodal communication with visual cues and more elaborated inter-speaker
adaptation. In addition, syllable counting is most of the time a straightforward task in one’s
mother language, even if the determination of syllable boundaries themselves may be
ambiguous. On the contrary, segment counting is well-known to be prone to variation and
inconsistency (see Port and Leary (2005): 941 inter alia). Beside the methodological advantage
of syllable for counting, numerous studies suggested its role either as a cognitive unit or as a unit
of organization in speech production or perception (e.g. Schiller (2008); Segui and Ferrand
(2002); but see Ohala (2008)). Hence, following Ladefoged (1975), we consider that ‘a syllable
is a unit in the organization of the sounds of an utterance’ (Ladefoged 2007) and, as far as the
distribution of linguistic information is concerned, it seems reasonable to investigate whether
syllabic speech rate really varies from one language to another and to what extent it influences
the speech information rate.

The MULTEXT corpus used in the present study was not gathered for this purpose, but it
provides a useful resource to address this issue, because of the similar content and recording
conditions across languages. We thus carried out measurements of the speech rate in terms of the
number of syllables per second for each recording of each speaker (the Syllable Rate, SR).
Moreover, the gross mean values of SR among individuals and passages were also estimated for
each language ($SR_L$, see Figure 1).

In parallel the 585 recordings were used to fit a model to SR using the linear mixed-
model procedure with Language and Speaker’s Sex as independent (fixed effect) predictors and
Speaker identity and Text as independent random effects. Note that in all the regression analyses
reported in the rest of this paper, a z-score transformation was applied to the numeric data, in order to get effect estimates of comparable magnitudes.

A preliminary visual inspection of the q-q plot of the model’s residuals led to the exclusion of 15 outliers whose standardized residuals were distant from zero by more than 2.5 standard deviations. The analysis was then rerun with the 570 remaining recordings and the visual inspection showed no longer deviation from normality, confirming that the procedure was suitable. We observed a main effect of Language, with highly significant differences among most of the languages: all $p_{\text{MCMC}}$ were inferior to .001 except between English and German ($p_{\text{MCMC}} = .08$, ns), French and Italian ($p_{\text{MCMC}} = .55$, ns), and Japanese and Spanish ($p_{\text{MCMC}} = .32$, ns). There is also a main effect of Sex ($p_{\text{MCMC}} = .0001$), with higher SR for male speakers than for female speakers, which is consistent with previous studies (e.g. Jacewicz, et al. (2009); Verhoeven, De Pauw, and Kloots (2004)).

Both Text ($\chi^2(1) = 269.79, p < .0001$) and Speaker ($\chi^2(1) = 684.96, p < .0001$) were confirmed as relevant random-effect factors, as supported by the likelihood ratio analysis, and kept in subsequent analyses.

**INSERT FIGURE 1 HERE**

The presence of different oral styles in the corpus design (narrative texts and queries) is likely to influence SR (Kowal et al. 1983), and thus explains the main random effect of Text. Besides, the main fixed effect of Language supports the idea that languages make different use of the temporal dimension during speech communication. Consequently, SR can be seen as
resulting from several factors: Language, nature of the production task and variation due to the Speaker (including the physiological or sociolinguistic effect of Sex).

3.4. A REGULATION OF THE SPEECH INFORMATION RATE. We investigated here the possible correlation between $ID_L$ and $SR_L$, with a regression analysis on SR, again using the linear mixed-model technique. ID is now considered in the model as a numerical covariate, beside the factors taken into account in the previous section (Language, Sex, Speaker, and Text).

We observed a highly significant effect of ID ($p_{MCMC} = .0001$) corresponding to a negative slope in the regression. The estimated $\beta$ value for the effect of ID is $\beta = -0.137$ with a 95% confidence interval in the range $[-0.194, -0.084]$. This significant regression demonstrates that the languages of our sample exhibit regulation, or at least a relationship, between their linguistic encoding and their speech rate.

Consequently, it is worth examining the overall quantity of information conveyed by each language per unit of time (and not per syllable). This so-called Information Rate ($IR$) encompasses both the strategy of linguistic encoding and the speech settings for each language $L$.

Again $IR$ is calculated using VI as an external point of reference:

$$IR_k(spkr) = \frac{S_k}{D_k(spkr)} \times \frac{\overline{D_k(VI)}}{S_k} = \frac{\overline{D_k(VI)}}{D_k(spkr)}$$

where $D_k(spkr)$ is the duration of the Text number $k$ uttered by speaker $spkr$. Since there is no a priori motivation to match one specific speaker $spkr$ of language $L$ to a given speaker of VI, we used the mean duration for text $k$ in Vietnamese $\overline{D_k(VI)}$. It follows that $IR_k$ is superior to one when the speaker has a higher syllabic rate than the average syllabic rate in Vietnamese for the same text $k$. Next, $IR_L$ corresponds to the average amount of information conveyed per unit of
time in language $L$ and it is defined as the mean value of $IR_\text{avg}(\text{spkr})$ among the speakers of language $L$.

Each speaker can definitely modulate her own speech rate to deviate from the average value as a consequence of the sociolinguistic and interactional context. However, our hypothesis is that the language identity would not be an efficient predictor of $IR$ because of functional equivalence among languages. To test this hypothesis, we fitted a model to $IR$ using the linear mixed-model procedure with Language and Speaker’s Sex as independent (fixed effect) predictors and Speaker identity and Text as independent random effects. Again, both Text ($\chi^2(1) = 414.75, p < .0001$) and Speaker ($\chi^2(1) = 176.55, p < .0001$) were confirmed as relevant random-effect factors. Sex was also identified as a significant fixed-effect predictor ($p_{\text{MCMC}} = .002$) and, contrary to our prediction, the contribution of Language was also significant for pairs involving Japanese and English. More precisely, JA contrasts with the 6 other languages ($p_{\text{MCMC}} < .001$ for all pairs), and EN significantly contrasts with JA, GE, MA, SP ($p_{\text{MCMC}} < .01$) and with IT and FR ($p_{\text{MCMC}} < .05$). Our hypothesis of equal IR among languages is thus invalidated, even if 5 of the 7 languages cluster together (GE, MA, IT, SP, and FR).

Figure 2 displays on the same graph $ID_L$, $SR_L$ and $IR_L$. For convenience, $ID_L$, which is unitless, has been multiplied by 10 to be represented on the same scale as $SR_L$ (left axis). The interaction between Information Density (grey bars) and Speech Rate (dashed bars) is visible since the first one increases while the second one decreases. The black dotted line connects the Information Rates values for each language (triangle marks, right axis). English ($IR_{EN} = 1.08$) shows a higher Information Rate than Vietnamese ($IR_{VI} = 1$). On the contrary, Japanese exhibits the lowest $IR_L$ value of the sample. Moreover, one can observe that several languages may reach
very close $IR_L$ with different encoding strategies: Spanish is characterized by a fast rate of low-density syllables while Mandarin exhibits a 34% slower syllabic rate with syllables ‘denser’ by a factor of 49%. Finally, their Information Rates differ only by 4%.

**INSERT FIGURE 2 HERE**

3.5. **SYLLABIC COMPLEXITY AND INFORMATION TRADE-OFF.** Among possible explanations of the density/rate trade-off, it may be put forward that if the amount of information carried by syllables is proportional to their complexity (defined as their number of constituents), the information density is positively related to the average syllable duration and consequently negatively correlated to syllable rate. We consider this line of explanation is this section.

**TOWARDS A MEASURE OF SYLLABIC COMPLEXITY.** A traditional way to estimate phonological complexity is to evaluate the size of the repertoire of phonemes for each language (Nettle 1995), but this index is unable to cope with the sequential language-specific constraints existing between adjacent segments or at the syllable level. It has therefore been proposed to directly consider this syllable level either by estimating the size of the syllable inventory (Shosted (2006)) or by counting the number of phonemes constituting the syllables (Maddieson, 2006). This latter index has also been extensively used in psycholinguistic works investigating the potential relationship between linguistic complexity and working memory (e.g., Mueller et al. (2003); Service (1998)), but it ignores all non-segmental phonological information such as tone, though it carries a very significant part of the information (Surendran and Levow 2004).

Here, we define the syllable complexity as its number of *constituents* (both segmental and tonal). In the 7-language dataset, only Mandarin requires taking the tonal constituent into account and the complexity of each of its tone bearing syllables is thus computed by adding 1 to
its number of phonemes while for neutral tone syllables, the complexity is equal to the number of phonemes. This is also the case for the syllables of the 6 other languages. An average syllabic complexity for a language may therefore be computed given an inventory of the syllables occurring in a large corpus of this language. Since numerous studies in linguistics (Bybee (2006); Hay et al. (2001); inter alia) and psycholinguistics (e.g., see Levelt (2001); Cholin et al. (2006)) point toward the relevance of the notion of frequency of use, we computed this syllabic complexity index both in terms of type (each distinct syllable counting once in the calculation of the average complexity) and token (the complexity of each distinct syllable is weighted by its frequency of occurrence in the corpus).

These indices were computed from large syllabified written corpora gathered for psycholinguistic purposes. Data for French are derived from the LEXIQUE 3 database (New et al. 2004); data for English and German are extracted from WebCelex database (Baayen et al. 1993); data for Spanish and Italian come from (Pone 2005); data for Mandarin are extracted from (Peng 2005) and converted into pinyin using the Chinese Word Processor (© NJStar Software Corp.); and data for Japanese are computed from (Tamaoka and Makioka 2004).

Data are displayed in Table 2. For each language, the second column gives the size of the syllable inventories, and the third and fourth columns give the average syllabic complexity for types and tokens, respectively. On average, the number of constituents estimated from tokens is 0.95 smaller than the number of constituents estimated from types. Leaving the tonal constituent of Mandarin syllables aside, this confirms the notorious fact that shorter syllables are more frequent than longer ones in language use. Japanese exhibits the lowest syllabic complexity – whether per types or per tokens - while Mandarin reaches the highest values.
SYLLABIC COMPLEXITY AND INFORMATION. The paradigmatic measures of syllabic complexity, for types and tokens, echo the syntagmatic measure of information density previously estimated on the same units – syllables. It is thus especially relevant to evaluate whether the trade-off revealed in Section 3.4 is due to a direct relation between the syllabic information density and the syllable complexity (in terms of number of constituents).

In order to assess whether the indices of syllabic complexity are related to the density/rate trade-off, their correlations with Information Density, Speech Rate and Information Rate were computed. The very small size of the sample \( n = 7 \) languages strongly limits the reliability of the results, but nevertheless gives an insight on future research directions. For the same reason, we estimated the correlation according to both Pearson’s correlation analysis \( (r) \) and Spearman’s rank correlation analysis \( (\rho) \), in order to potentially detect incongruent results. Eventually, both measures of syllabic complexity are correlated to \( ID_L \). The highest correlation is reached with the complexity estimated on the syllable types \( (\rho = 0.98, p < .01; r = 0.94, p < .01) \), and is further illustrated in Figure 3. Speech Rate \( SR_L \) is negatively correlated to syllable complexity estimated from both types \( (\rho = -0.98, p < .001; r = -0.83, p < .05) \) and tokens \( (\rho = -0.89, p < .05; r = -0.87, p < .05) \). These results suggest that syllable complexity is engaged in a twofold relationship with the two terms of the trade-off highlighted on information density rate \( (ID_L \text{ and } SR_L) \) without enabling one to disentangle the causes and consequences of this scheme. Furthermore, an important result is that no significant correlation is evidenced between the syllabic information rate and indices of syllabic complexity, both in terms of type \( (\rho = 0.16, p =0.73; r = 0.72, p=.07) \) and token \( (\rho = 0.03, p=0.96; r = 0.24, p =0.59) \).
4. DISCUSSION

4.1. TRADE-OFF BETWEEN INFORMATION DENSITY AND SPEECH RATE. Two hypotheses motivate the approach taken in this paper. The first one states that, for functional reasons, the rate of linguistic information transmitted during speech communication is, to some extent, similar among languages. The second hypothesis is that this regulation results in a density/rate trade-off between the average information density carried by speech chunks and the number of chunks transmitted per second.

Regarding Information Density, results show that the seven languages of the sample exhibit a large variability especially highlighted by the ratio of one-half between Japanese and Vietnamese $ID_L$. Such variation is related to language-specific strategies, not only in terms of pragmatics and grammars (what is explicitly coded and how) but also in terms of word formation rules, which, in turn, may be related to syllable complexity (see Fenk-Oczlon and Fenk (1999); Plotkin and Novack (2000)). One may object that in Japanese morae are probably more salient units than syllables to account for linguistic encoding. Nevertheless, syllables give ground to a methodology that can be strictly transposed from one language to another, as far as average information density is concerned.

Syllabic speech rate significantly varies as well, both among speakers of the same language and cross-linguistically. Since the corpus was not explicitly recorded to investigate
speech rate, the variation observed is an underestimation of the potential range from slow to fast speech rates, but it provides a first approximation of the ‘normal’ speech rate by simulating social interactions. Sociolinguistic arguments pointing to systematic differences in speech rates among populations speaking different languages are not frequent, one exception being the hypothesis that links a higher incidence of fast speech to small, isolated communities (Trudgill 2004). Additionally, sociolinguists often consider that, within a population, speech rate is a factor connected to speech style and is involved in a complex pattern of status and context of communication (Brown, et al. (1985); Wells (1982)).

Information rate is shown to result from a density/rate trade-off illustrated by a very strong negative correlation between the $ID_L$ and $SR_L$. This result confirms the hypothesis suggested fifty years ago by Karlgren (1961:676) and reactivated more recently (Greenberg and Fosler-Lussier (2000); Locke (2008)): ‘It is a challenging thought that general optimalization rules could be formulated for the relation between speech rate variation and the statistical structure of a language. Judging from my experiments, there are reasons to believe that there is an equilibrium between information value on the one hand and duration and similar qualities of the realization on the other’ (Karlgren 1961). However, $IR_L$ exhibits more than 30% of variation between Japanese (0.74) and English (1.08), invalidating the first hypothesis of a strict cross-language equality of rates of information. The linear mixed-effect model nevertheless reveals that no significant contrast exists among 5 of the 7 languages (GE, MA, IT, SP, and FR), and highlights that texts themselves and speakers are very significant sources of variation. Consequently, one has to consider the alternative loose hypothesis that $IR_L$ varies within a range of values that guarantee an efficient communication, fast enough to convey useful information and slow enough to limit the communication cost (in its articulatory, perceptual and cognitive
dimensions). However, a deviation from the optimal range of variation defined by these constraints is still possible because of additional factors, such as social aspects. A large-scale study involving many languages would eventually confirm or invalidate the robustness of this hypothesis, answering to questions such as: what is the possible range of variation for ‘normal’ speech rate and information density? To what extent can a language depart from the density/rate trade-off? Can we find a language with both a high speech rate and a high information density?

These results support the idea that, despite the large variation observed in phonological complexity among languages, a trend toward regulation of the information rate is at work, as illustrated here by Mandarin and Spanish reaching almost the same average information rate with two opposite strategies: slower, denser and more complex for Mandarin vs. faster, less dense and less complex for Spanish. The existence of this density/rate trade-off may thus illustrate a twofold least-effort equilibrium in terms of ease of information encoding and decoding in the one hand, vs. efficiency of information transfer through the speech channel, in the other.

In order to provide a first insight on the potential relationship between the syntagmatic constraints on information rate and the paradigmatic constraints on syllable formation, we introduced type-based and token-based indices of syllabic complexity. Both are positively correlated to information density and negatively correlated to syllabic rate. However, one has to be cautious with these results for at least two reasons. The first one is that the language sample is very small which leads to results that have no typological range (this caveat is obviously valid for all the results presented in this article). The second shortcoming is due to the necessary counting of the constituents for each syllable, leading to questionable methodological choices mentioned earlier for phonemic segmentation and for weighting tone and stress dimensions.
4.2. INFORMATION-DRIVEN REGULATIONS AT THE INDIVIDUAL LEVEL. Another noticeable result is that the syllabic complexity indices do not correlate with the observed Information Rate $IR_L$. Thus, these linguistic factors of phonological complexity are bad – or at least insufficient – predictors of the rate of information transmission during speech communication. These results bring new cross-linguistic arguments in favour of a regulation of the information flow.

This study echoes recent works investigating informational constraints on human communication (Aylett and Turk, (2004); Frank and Jaeger (2008); Genzel and Charniak (2002); Keller (2004); Levy and Jaeger (2007)), with the difference that it provides a cross-language perspective on the average information rather than a detailed language-specific study of the distribution of information (see also van Son and Pols (2003)). All these studies have in common the assumption that human communication may be analysed through the prism of Information Theory, and that humans try to optimally use the channel of transmission through a principle of Uniform Information Density (UID). This principle postulates ‘that speakers would optimize the chance of successfully transmitting their message by transmitting a uniform amount of information per transmission (or per time, assuming continuous transmission) close to the Shannon capacity of the channel.’ (Frank and Jaeger 2008). The authors postulated that speakers would try avoiding spikes in the rate of information transmission in order to avoid ‘wasting’ of some channel capacity. However, this hypothesis is controversial, especially because what is optimal from Shannon’s theory (transmission) is not necessarily optimal for human cognitive processing (coding and decoding). It is thus probable that the transmission constraints also interact with other dimensions such as probabilistic paradigmatic relations, as suggested in Kuperman et al. (2007), or attentional mechanisms, for instance.
More generally, information-driven trade-offs could reflect general characteristics of information processing by human beings. Along these lines, frequency matching and phase locking between the speech rate and the activations in the auditory cortex during a task of speech comprehension (Ahissar et al. 2001) would be worth investigating in a cross-language perspective to elucidate whether these synchronizations are also sensitive to the information rate for the considered languages. Working memory refers to the structures and processes involved in the temporary storage and processing of information in the human brain. One of the most influential models includes a system called the phonological loop (Baddeley (2000, 2003); Baddeley and Hitch (1974)) that would enable one to keep a limited amount of verbal information available to the working memory. The existence of time decay in memory span seems plausible (see Schweickert & Boruff (1986) for a mathematical model) and several factors may influence the capacity of the phonological loop. Among others, articulation duration, phonological similarity, phonological length (viz. the number of syllables per item) and phonological complexity (viz. the number of phonological segments per item) are often mentioned (Baddeley (2000); Mueller et al. (2003); Service (1998)). Surprisingly, mother tongues of the subjects and languages of the stimuli have not been thoroughly investigated per se as relevant factors. Tasks performed in spoken English vs. American Sign Language have indeed revealed differences (Boutla et al. (2004); Bavelier et al. (2006)), but without determining for certain whether they were due to the distinct modalities, to the distinct linguistic structures or to neurocognitive differences in phonological processing across populations. However, since there are significant variations in speech rate among languages, cross-language experiments would probably provide major clues for disentangling the relative influence of universal general processes vs. linguistic parameters. If one assumes constant time decay for the memory span, it
would include a different number of syllables according to different language-specific speech rates. On the contrary, if linguistic factors matter, one could imagine that the differences across languages in terms of syllabic complexity or speech rate would influence memory spans. As a conclusion, we would like to point out that cross-language studies may be very fruitful for revealing whether the memory span is a matter of syllables, words, quantity of information or simply duration. More generally, such cross-language studies are crucial both for linguistic typology and for language cognition (see also Evans and Levinson (2009)).
5. REFERENCES


Bell, Alan, Jason M. Brenier, Michelle Gregory, Cynthia Girand, and Dan Jurafsky. 2009. Predictability effects on durations of content and function words in conversational English. Journal of Memory and Language 60:1.92-111


R Development Core Team. 2010. R: A language and environment for statistical computing. R
Foundation for Statistical Computing, Vienna, Austria. ISBN 3-900051-07-0, URL

Roach, Peter. 1999. Some languages are spoken more quickly than others. Language myths, ed.


Schiller, Niels O. 2008 Syllables in Psycholinguistic Theory: Now you see them, Now You
Don’t. Syllable in Speech Production, ed. by B. L. Davis and K. Sajdó, 155-176. New-York:
Lawrence Erlbaum Associates.

Schweickert, Richard, and Boruff, B. 1986. Short-term memory capacity: magic number or
magic spell. Journal of Experimental Psychology, Learning, Memory, and Cognition 12.419-
425.

Segui, Juan, and Ferrand, L. 2002. The role of the syllable in speech perception and production.
University Press.

Service, Elisabet. 1998. The effect of word-length on immediate recall depends on phonological
complexity, not articulation duration. Quaterly Journal of Experimental Psychology 51:A.283-
304.

Shannon, Claude E., and Weaver, W. 1949. The mathematical theory of information. Urbana:
University of Illinois Press.

10:1.1-40.


See King, (1967) for an overview of the genesis of this notion.

This influence is especially visible in the fact that Hockett considered redundancy as the first of its phonological universals: ‘In every human language, redundancy, measured in phonological terms, hovers near 50%.’ (Hockett, 1966: 24), with an explicit reference to Shannon.

This filtering was done because in each language, pause durations widely vary from one speaker to another, probably because no explicit instructions were made to the speakers.

See Baayen et al. (2008) for a detailed description of this technique and a comparison to other statistical analyses. More generally, all statistical analyses were done with R 2.11.1 (R Development Core Team, 2010). The mixed-effect model was estimated using the lme4 and languageR packages. Pearson’s correlations are obtained with the cor.test function (standard parameters). Because of the very small size of the language sample (n = 7), Spearman rho are computed using the spearman.test function from the pspearman package, which uses the exact null distribution for the hypothesis test with small samples (n < 22).

The speaker’s Sex was actually taken into consideration: the normalization was based on the mean duration among either Vietnamese female speakers or Vietnamese male speakers, depending on spkr sex.

Further investigations would actually be necessary to exactly quantify the average weight of the tone in the syllable complexity. The value ‘1’ simply assumes an equal weight for tonal and segmental constituents in Mandarin. Furthermore, lexical stress (e.g. in English) could also be considered as a constituent but to the best of the authors’ knowledge, its functional load (independently from the segmental content of the syllables) is not precisely known.
English: Last night I opened the front door to let the cat out. It was such a beautiful evening that I wandered down the garden for a breath of fresh air. Then I heard a click as the door closed behind me. I realised I'd locked myself out. To cap it all, I was arrested while I was trying to force the door open!

French: Hier soir, j'ai ouvert la porte d'entrée pour laisser sortir le chat. La nuit était si belle que je suis descendu dans la rue prendre le frais. J'avais à peine fait quelque pas que j'ai entendu la porte claquer derrière moi. J'ai réalisé, tout d'un coup, que j'étais fermé dehors. Le comble c'est que je me suis fait arrêter alors que j'essayais de forcer ma propre porte !

Italian: Ieri sera ho aperto la porta per far uscire il gatto. Era una serata bellissima e mi veniva voglia di starmene sdraiato fuori al fresco. All'improvviso ho sentito un clic dietro di me e ho realizzato che la porta si era chiusa lasciandomi fuori. Per concludere, mi hanno arrestato mentre cercavo di forzare la porta!

Japanese: 昨夜、私は猫を外に出してやるために玄関を開けてみると、あまりに気持ちのいい夜だったので、新鮮な空気をす吸おうと、ついふらっと庭へ降りたのです。すると後ろでドアが閉まって、カチッと言う音が聞こえ、自分自身を締め出してしまったことに気が付いたのです。挙げ句の果てに、私は無理矢理ドアをこじ開けようとしているところを逮捕されてしまったのです。


Mandarin Chinese: 昨夜我打开前门放猫出去的时候，看到夜色很美，就走下台阶，想到花园里呼吸呼吸新鲜空气。当时只听到身后咔哒一声，发现自己被锁在门外了。更糟的是，当我试图撬开门的时候被警察逮捕了。

Spanish: Anoche, abrí la puerta del jardín para sacar el gato. Hacia una noche tan buena que pensé en dar un paseo y respirar el aire fresco. De repente, se me cerró la puerta. Me quedé en la calle, sin llaves. Para rematarlo, me arrestaron cuando trataba de forzar la puerta para entrar.

APPENDIX : TRANSLATIONS OF TEXT P8
<table>
<thead>
<tr>
<th>Language</th>
<th>Informational Density ($ID_L$)</th>
<th>Syllable Rate (#syl / s)</th>
<th>Information Rate</th>
</tr>
</thead>
<tbody>
<tr>
<td>English</td>
<td>0.91 (± 0.04)</td>
<td>6.19 (± 0.16)</td>
<td>1.08 (± 0.08)</td>
</tr>
<tr>
<td>French</td>
<td>0.74 (± 0.04)</td>
<td>7.18 (± 0.12)</td>
<td>0.99 (± 0.09)</td>
</tr>
<tr>
<td>German</td>
<td>0.79 (± 0.03)</td>
<td>5.97 (± 0.19)</td>
<td>0.90 (± 0.07)</td>
</tr>
<tr>
<td>Italian</td>
<td>0.72 (± 0.04)</td>
<td>6.99 (± 0.23)</td>
<td>0.96 (± 0.10)</td>
</tr>
<tr>
<td>Japanese</td>
<td>0.49 (± 0.02)</td>
<td>7.84 (± 0.09)</td>
<td>0.74 (± 0.06)</td>
</tr>
<tr>
<td>Mandarin</td>
<td>0.94 (± 0.04)</td>
<td>5.18 (± 0.15)</td>
<td>0.94 (± 0.08)</td>
</tr>
<tr>
<td>Spanish</td>
<td>0.63 (± 0.02)</td>
<td>7.82 (± 0.16)</td>
<td>0.98 (± 0.07)</td>
</tr>
<tr>
<td>Vietnamese</td>
<td>1 (reference)</td>
<td>5.22 (± 0.08)</td>
<td>1 (reference)</td>
</tr>
</tbody>
</table>

**Table 1.**

Cross-language comparison of Informational Density, Syllabic Rate and Information Rate (mean values and 95% confidence intervals). Vietnamese is used as the external reference.
<table>
<thead>
<tr>
<th>LANGUAGE</th>
<th>SYLLABLE INVENTORY SIZE N</th>
<th>SYLLABLE COMPLEXITY (TYPE) AVERAGE #CONSTITUENTS/SYL</th>
<th>SYLLABLE COMPLEXITY (TOKEN) AVERAGE #CONSTITUENTS/SYL</th>
</tr>
</thead>
<tbody>
<tr>
<td>English</td>
<td>7,931</td>
<td>3.70</td>
<td>2.48</td>
</tr>
<tr>
<td>French</td>
<td>5,646</td>
<td>3.50</td>
<td>2.21</td>
</tr>
<tr>
<td>German</td>
<td>4,207</td>
<td>3.70</td>
<td>2.68</td>
</tr>
<tr>
<td>Italian</td>
<td>2,719</td>
<td>3.50</td>
<td>2.30</td>
</tr>
<tr>
<td>Japanese</td>
<td>416</td>
<td>2.65</td>
<td>1.93</td>
</tr>
<tr>
<td>Mandarin</td>
<td>1,191</td>
<td>3.87</td>
<td>3.58</td>
</tr>
<tr>
<td>Spanish</td>
<td>1,593</td>
<td>3.30</td>
<td>2.40</td>
</tr>
</tbody>
</table>

Table 2.

Cross-language comparison of syllabic inventory, and syllable complexities (in terms of number of constituents).
Speech rates measured in terms of the number of syllables per second (mean values and 95% confidence intervals). Stars illustrate significant differences between the homogeneous subsets revealed by post hoc analysis.
Comparison of the encoding strategies of linguistic information. Grey and dashed bars respectively display Information Density $ID_L$ and Syllabic Rate $SR_L$ (left axis). For convenience, $ID_L$ has been multiplied by a factor of 10. Black triangles give Information Rate $IR_L$ (right axis, 95% confidence intervals displayed). Languages are ranked by increasing $ID_L$ from left to right (see text).
FIGURE 3.

Relation between Information Density and Syllable complexity (average number of constituents per syllable type). 95% confidence intervals are displayed.