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The forms and meanings of grammatical markers support efficient communication

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Abstract

Functionalist accounts of language suggest that forms are paired with meanings in ways that support efficient communication. Previous work on grammatical marking suggests that word forms have lengths that enable efficient production, and work on the semantic typology of the lexicon suggests that word meanings represent efficient partitions of semantic space. Here we establish a theoretical link between these two lines of work and present an information-theoretic analysis that captures how communicative pressures influence both form and meaning. We apply our approach to the grammatical features of number, tense, and evidentiality, and show that the approach explains both which systems of feature values are attested across languages and the relative lengths of the forms for those feature values. Our approach shows that general information-theoretic principles can capture variation in both form and meaning across languages.

1 Introduction

A primary goal of linguistic typology is to characterize and explain the diversity in extant linguistic systems compared to possible but unattested systems (Croft, 2002). Linguistic typology can be approached from a variety of perspectives (e.g., Newmeyer, 2005), but here we take a functional approach and build on a large body of work that has explored ways in which language supports efficient communication (von der Gabelentz, 1901; Zipf, 1949; J. Bybee, 2010; Hawkins, 2004, 2014). Recent work in this tradition has formalized communicative efficiency in terms of information theory and has used this formalization to demonstrate that linguistic forms and meanings support efficient communication (Gibson et al., 2019), but form and meaning are usually treated separately. On one hand, a substantial body of work has demonstrated that linguistic forms allow communication with a minimum of effort (Zipf, 1949; Greenberg, 1966; Piantadosi, Tily, & Gibson, 2011; Haspelmath, 2021), but this work typically does not explain the meanings associated with the forms in question. On the other hand, recent work in semantic typology has shown that word meanings within several semantic domains support efficient communication (Regier, Kemp, & Kay, 2015; Kemp, Xu, & Regier, 2018), but does not address the forms used to express these meanings. Here we show that an existing information-theoretic account of lexical semantics (Kemp et al., 2018, as formulated in Zaslavsky, Kemp, Regier, & Tishby, 2018) also accounts for classic ideas about coding efficiency from the literature on grammatical marking (Greenberg, 1966; Haspelmath, 2021). Connecting these lines of work illustrates how information theory provides a unified account of both the meanings encoded in natural language and the forms used to express them.

Our theoretical framework applies to both grammar and the lexicon, but we focus here on grammatical marking expressed by morphology, and in particular on the grammatical features of number, tense and evidentiality. We chose these features because they primarily convey semantic information and because each encodes a rich semantic dimension instead of a simple binary distinction. Number reflects the number of entities involved in one role of an event (e.g., four lions chasing a giraffe). Tense refers to the location of an event in time (e.g., past, present, future). Evidentiality refers to the source of information (e.g., did I see it, or hear someone else describe it?). Grammatical features like these are core components of language, yet there is considerable variation in the size of grammatical feature inventories and the realization of grammatical features across languages. For example, the data analyzed in this paper include fifteen distinct morphological systems that languages use to mark grammatical number. Whereas English only distinguishes between singular and plural, Larike distinguishes between singular, dual, trial and plural (Laidig & Laidig,

1990). Accounting for the diversity of feature inventories and realizations across languages is therefore a significant challenge.

Our work builds on functionalist accounts of grammatical features from several areas of the literature. A longstanding line of work has used corpus analyses to show that the realizations of grammatical feature values are shaped by the principle of least effort (Zipf, 1949). Because speakers often need to convey the meanings associated with grammatical features, grammatical markers have short forms that are easy to produce, and the most frequent feature values may receive no overt marking (Greenberg, 1966; Haspelmath & Karjus, 2017; Haspelmath, 2021). A second line of work has used artificial language learning experiments and evolutionary models to demonstrate that learners restructure their input to produce systems that are simpler, easier to produce, and more informative (Fedzechkina, Jaeger, & Newport, 2012; Kanwal, Smith, Culbertson, & Kirby, 2017; Kurumada & Grimm, 2019; Fedzechkina & Jaeger, 2020), often in line with linguistic universals (e.g., Culbertson, Smolensky, & Legendre, 2012). Related work has also demonstrated that more easily acquired grammatical systems occur more frequently in the world's languages (Gentner & Bowerman, 2009; Saratsli, Bartell, & Papafragou, 2020). Our approach is broadly consistent with all of these research strands, but formally bridges them by providing an integrated account of both grammatical feature values and the forms used to express them.

The information-theoretic framework that we use formalizes the trade-off between informativeness and simplicity that languages must negotiate. Consider a speaker who wishes to convey some meaning (e.g., the number of empty coffee cups still on my desk) to a hearer (see Figure 1 top panel). A highly informative system allows the speaker to discriminate between many different meanings (e.g., many different numbers of cups), but this communicative precision can only be achieved if the system is far from simple. The trade-off between informativeness and simplicity has been discussed for many years in the literature on "competing motivations" (Haiman, 2010; von der Gabelentz, 1901; Du Bois, 1985) and several measures of morphological simplicity have been proposed (see Supplemental Information for a discussion). Here, we build on a recent account of lexical semantics (Zaslavsky et al., 2018; Zaslavsky, 2020) that is grounded in Rate–Distortion theory (Shannon, 1948, 1959, the branch of information theory characterizing efficient data compression), and that formalizes both informativeness and simplicity in information-theoretic terms. Within this framework, the simplicity dimension connects naturally with the notion of coding efficiency from the literature on grammatical marking (Hawkins, 2004; Haspelmath & Karjus, 2017). We will therefore argue that the trade-off between informativeness and simplicity helps to explain both which feature values are attested across languages and the relative lengths of the linguistic realizations of these feature values.

The next section introduces our theoretical framework and provides formal definitions of information loss (the inverse of informativeness) and complexity (the inverse of simplicity). We then provide an overview of number, tense and evidentiality across languages and introduce the typological data that we analyze. The first set of analyses focuses on meaning, and demonstrates that grammatical feature inventories achieve near-optimal trade-offs between informativeness and simplicity. The second set of analyses focuses on form, and demonstrates that the realizations of grammatical features enable concise communication.

2 Theoretical Framework

We build on the theoretical framework in Zaslavsky et al. (2018), which has been previously used to account for word meanings across languages, and show that the same framework can also be linked to aspects of linguistic form. The framework, illustrated in the top panel of Figure 1, assumes a speaker and a listener who wish to communicate about states of the world u drawn from the universe $\mathcal{U}$. The speaker is uncertain about the true state of the world, and their mental state is captured by a speaker distribution s over states in $\mathcal{U}$. To summarize this mental state the speaker generates a linguistic form f according to an encoder $q(f|s)$ which maps speaker distributions into forms. Upon receiving this form, the listener computes a distribution $\hat{s}$ that is intended to approximate the speaker distribution s. We assume that this distribution $\hat{s}$ is computed by carrying out Bayesian inference based on the encoder $q(f|s)$ and a prior $p(s)$ over speaker distributions, which gives the optimal $\hat{s}$ (Zaslavsky et al., 2018). The prior reflects communicative need, or the relative frequency with which speakers communicate about different states of the world (Gibson et al., 2019; Zaslavsky, Kemp, Tishby, & Regier, 2019; Karjus, Blythe, Kirby, & Smith, 2020).

An optimal encoder $q(f|s)$ should satisfy two criteria: it should allow the listener to accurately reconstruct

the speaker's mental state, and it should minimize production effort by ensuring that frequently-used forms are short. To formalize these criteria it will be convenient to represent a grammatical marker as a pair (m, f) that includes both a meaning (or feature value) m and a form (or realization) f. For number in English there are two such pairs: (SINGULAR, $\emptyset$) and (PLURAL, "-s"), where the empty set $\emptyset$ indicates that the singular is zero marked, or realized without an overt form. Given this representation we can decompose the encoder $q(f|s)$ into a meaning encoder $q_m(m|s)$ that maps speaker distributions into meanings and a form encoder $q_f(f|m)$ that maps meanings into forms. This two-stage encoding process is illustrated in Figure 1, and is used in the following sections to develop analyses that focus on efficiency of meaning and analyses that focus on efficiency of form.

The meaning encoder q_m is lossy, but for simplicity we assume that the form encoder q_f is lossless, which means that the listener is able to reconstruct the meaning m without error given the form f. In reality this assumption does not hold. Languages permit ambiguity in the linguistic signal, and ambiguity (including ambiguity arising from reanalysis) may have implications for the historical emergence of grammatical forms (Traugott, 2011). Assuming that q_f is lossless, however, is a natural starting point given the cross-linguistic data available to us.

2.1 Efficiency of meaning

An efficient encoder q_m achieves an optimal tradeoff between complexity and information loss (the inverse of informativeness). Following Zaslavsky et al. (2018), the formal definitions of complexity and information loss are grounded in the Information Bottleneck (IB) principle (Tishby, Pereira, & Bialek, 1999), which is a special type of a Rate–Distortion tradeoff. The complexity of an encoder measures how much information about the speaker's mental state is preserved in the meaning of a grammatical marker, and is defined as the mutual information between meanings and speaker distributions:

$$I(M; S) = H(M) - H(M|S), \tag{1}$$

which, as shown, can be formulated as the difference of two terms: the entropy over meanings $H(M)$ and the conditional entropy $H(M|S)$.

The informativeness of an encoder is negatively related to the expected information loss associated with each communicative interaction. Following Regier et al. (2015) and Zaslavsky et al. (2018) we define this information loss as the expected Kullback-Leibler divergence $\text{KL}(s||\hat{s})$ between the speaker distribution s and the listener's reconstruction $\hat{s}$ of that distribution:

$$E[\text{KL}[S||\hat{S}]] = \sum_{s,m} p(s) q_m(m|s) \text{KL}[s||\hat{s}_m], \tag{2}$$

where $\hat{s}_m$ is the reconstructed distribution for occasions on which the speaker chooses meaning m.

Every possible mapping from speaker distributions to meanings corresponds to a point in a two-dimensional space where the dimensions represent complexity and information loss. Some points in this space cannot be achieved by any possible language—for example, in any realistic setting it is impossible for an encoder to achieve both zero complexity and zero information loss. The boundary separating achievable points from unachievable points is a special case of a Pareto frontier known as the IB theoretical limit, and encoders along this continuous frontier achieve optimal trade-offs between complexity and information loss. These encoders are optimal in the sense that complexity cannot be reduced without increasing information loss, and information loss cannot be reduced without increasing complexity.

Given this theoretical framework, we can ask whether attested grammatical feature inventories achieve near-optimal trade-offs between complexity and information loss. For any given feature, applying the framework requires three components to be specified: the universe of world states $\mathcal{U}$, the speaker distributions for each objective world state s_u, and the prior on speaker distributions $p(s)$. Given these components we can compute the complexity and information loss of both attested and hypothetical systems, and trace out the IB Pareto frontier of systems that achieve optimal trade-offs between complexity and information loss (Tishby et al., 1999; Zaslavsky et al., 2018).
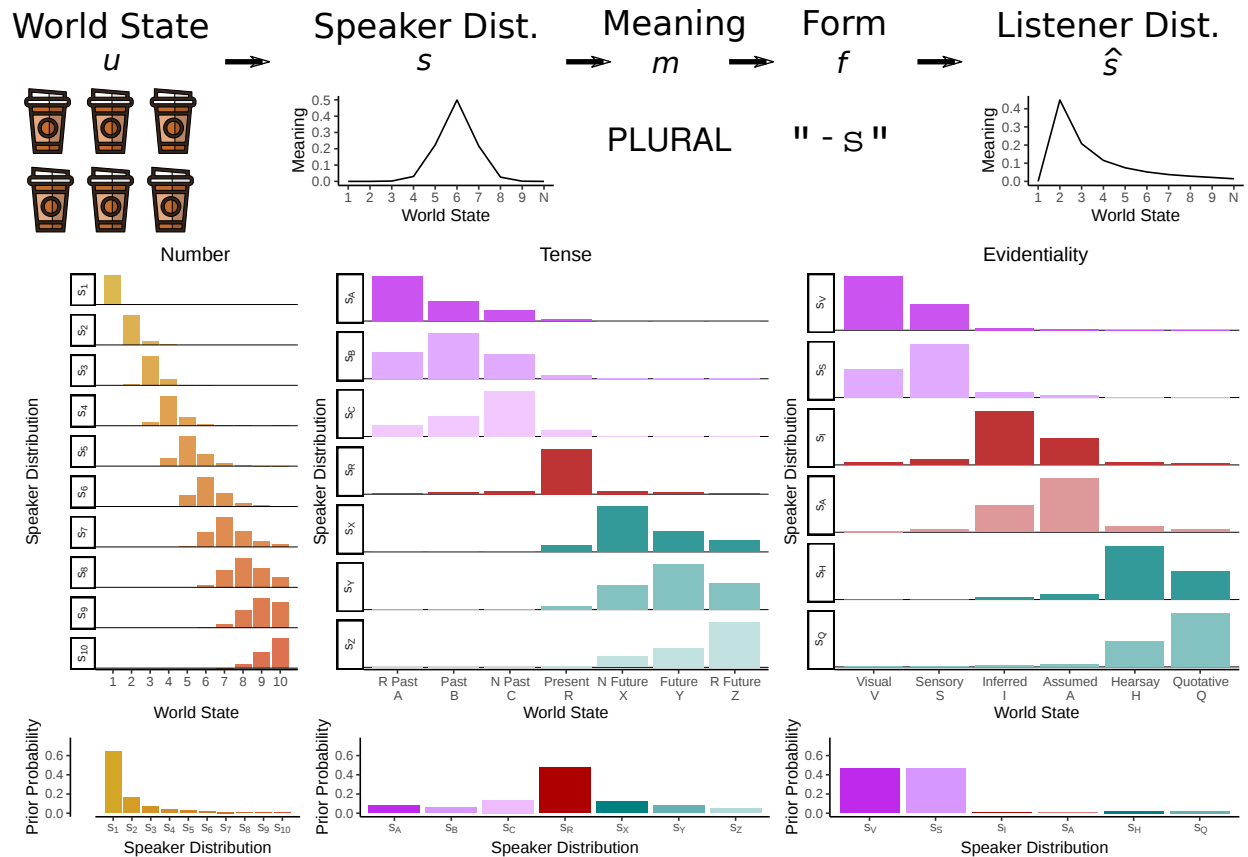
Figure 1: Communicative scenario along with speaker distributions and priors for number, tense and evidentiality. Top panels: Communicative scenario illustrating how a speaker generates a form which is then used by a listener to reconstruct the speaker distribution s over world states. In reality the form would not be uttered in isolation but rather combined with the noun "cup" to generate the utterance "cups." Center panels: Speaker distributions s_u for number, tense and evidentiality. Bottom panels: Priors $p(s)$ on the three sets of speaker distributions.

	Language	System	Complexity	Info. Loss	Front. Dist.	Form Corr.
Number	Pirahã	(GENERAL, ∅)	0.00	1.44	0.00	NA
	Russian	(SG, ∅)(PL, "ы")	0.94	0.55	0.01	1.00
	Larike	(SG, "mane")(DU$_O$, "matua")(TR$_O$, "matidu")(PL, "mati")	1.13	0.40	0.03	1.00
	Murrinh-Patha	(SG, "nukunu")(DU, "penintha")(PAUC, "peneme")(PL, "pigunu")	1.43	0.16	0.01	0.16
	Sursurunga	(SG, "i")(DU, "diar")(PAUC, "ditul")(GPAUC, "dihat")(PL, "di")	1.47	0.14	0.02	0.44
Tense	West Greenlandic	(ABCR, ∅)(XYZ, "ssa")	0.81	0.50	0.04	1.0
	Japanese	(ABC, "た")(RXYZ, ∅)	0.85	0.47	0.04	1.00
	Wolof	(ABC, "naa")(R, "nge")(XYZ, "dinaa")	1.52	0.08	0.00	1.00
	Hixkaryana	(A, "ye")(B, "yako")(C, "no")(RXYZ, "yaha")	1.26	0.42	0.21	0.32
	Zulu	(A, "a")(BC, "ile")(R, ∅)(X, "za")(YZ, "yaku")	2.01	0.02	0.00	0.88
Evidentiality	Sissala	(VSIA, ∅)(HQ, "ɛ")	0.28	0.15	0.00	1.0
	Abkhaz	(VS, ∅)(IAHQ, "заарен")	0.36	0.12	0.01	1.0
	Quechua	(VS, "mi")(IA, "chi")(HQ, "shi")	0.42	0.08	0.00	0.99
	Turkish	(V, ∅)(SIAHQ, "mış")	1.00	0.26	0.23	1.0
	Barasano	(V, "ka")(S, "ruyu")(IA, "ra")(HQ, "yu")	1.35	0.01	0.00	-0.09

Table 1: Example inventories for grammatical number, tense and evidentiality. Each system includes a single representative form for each meaning, and orthographic forms are shown instead of phonemic forms. The meanings for number are: GENERAL-"the noun can be expressed without reference to number" (Corbett, 2000, pg. 10); SG-Singular; PL-plural; DU-dual; TR-trial; PAUC-paucal (a few); GPAUC-greater paucal (a bunch); GPL-greater plural; and optional values are shown using the subscript $_O$. For tense, A, B and C denote distant, near, and immediate past, R denotes present, and X, Y and Z denote immediate, near and remote future. For evidentiality, V and S denote visual and sensory, I and A denote inferred and assumed, and H and Q denote hearsay and quotative. Frontier distance shows the Euclidean distance between a system and the corresponding Pareto frontier in Figure 2, and small values indicate efficiency of meaning. Form correlations show correlations between optimal and observed form lengths (see Figure 4) and large values indicate efficiency of form.

2.2 Efficiency of form

We now consider the mapping q_f from meanings to forms, or strings of phonemes. Because this mapping is assumed to be lossless, efficiency is purely a matter of minimizing expected form length. The entropy $H(M)$ gives a lower bound on expected form length (Shannon, 1948), and an efficient mapping q_f is expected to assign a form to meaning m that has length close to

$$h(m) = -\log p(m) = \log \sum_s q_m(m|s)p(s). \tag{3}$$

In reality, natural language mappings q_f are likely to yield expected codelengths that do not come especially close to the entropy $H(M)$ (Futrell, 2017; Pate, 2017; Pimentel, Nikkarinen, Mahowald, Cotterell, & Blasi, 2021). These mappings, however, may nevertheless reflect a pressure towards brevity. Equation 3 suggests that shorter forms should be used for more frequent meanings, and we will examine whether this inverse relationship between form length and frequency holds in our data.

Previous work on grammatical marking (Haspelmath, 2021; Fenk-Oczlon, 2001) and the lexicon (Zipf, 1949; Piantadosi et al., 2011; Bentz & Ferrer Cancho, 2016) has emphasized the notion of coding efficiency, and has demonstrated that forms tend to be paired with meanings in ways that allow utterances to be relatively concise. Similar results have emerged from studies of phonetic realization (Aylett & Turk, 2004), online word choice (Mahowald, Fedorenko, Piantadosi, & Gibson, 2013), and word ordering (Hahn, Jurafsky, & Futrell, 2020). Our theoretical approach is consistent with all of these results, but goes beyond them by considering efficiency of form within a framework that also captures efficiency of meaning.

Considering efficiency of form and meaning within a single framework is important because the two are connected via the entropy $H(M)$. Minimizing the expected codelength for an efficient code (i.e. minimizing $H(M)$) can only be achieved if the encoder q_m generates the same meaning for every speaker distribution. A system of this kind is maximally simple but also maximally uninformative (i.e. the information loss in Equation 2 is maximized). The pressure towards minimizing codelengths must therefore trade off against a pressure towards informative communication (Ferrer i Cancho & Solé, 2003). This tradeoff is especially clear for deterministic encoders q_m, for which the complexity measure in Equation 1 is equivalent to the entropy $H(M)$. Most (but not all) of the grammatical feature systems that we analyze are deterministic to a good first approximation: for example, in the English number system the singular is consistently used for individual items and the plural is consistently used for multiple items.

3 Framework instantiations

Now that we have introduced our general theoretical framework, we show how it can be applied to the grammatical features of number, tense and evidentiality. To evaluate our theory we will make several simplifying assumptions (Cooper & Guest, 2014; Guest & Martin, 2020). First, while number, tense and evidentiality can reflect multiple semantic dimensions (e.g., numerosity vs individuality, absolute vs relative time), we focus on a single grammaticalized semantic dimension for each. Each of these semantic dimensions can be expressed using a variety of strategies (e.g., numerals, adverbs, and verbal constructions), but for tractability we focus on systems that use obligatory morphological markers. As a result of these simplifications, the set of unique feature inventories in our sample is relatively small, and many languages are coded using a single feature value that spans the entire dimension. This kind of inventory is maximally simple (and therefore technically optimal), but also maximally uninformative. The languages coded in this way may not make use of obligatory grammatical markers, but typically rely on other linguistic constructions or contextual information for conveying information about the semantic dimensions in question. Evaluating the efficiency of these alternative communicative strategies is an important challenge for future work, and we return to this issue in the Discussion.

A second assumption is that the prior on speaker distributions $P(s)$ and the speaker distributions themselves are invariant across cultures. Previous studies have made similar assumptions (Kemp & Regier, 2012; Zaslavsky et al., 2018), and in all cases these assumptions should be viewed as rough first approximations that can be subsequently relaxed using data from studies that directly estimate culture-specific priors (Rácz, Passmore, Sheard, & Jordan, 2019). Third, our operationalization of production effort is relatively coarse, and we treat this quantity as a binary variable (marker present vs marker absent) or define it as the length of a marker's orthographic representation. Considering phonetic structure would allow for a more satisfying operationalization, but is not possible given the data available to us. Finally, the grammatical systems considered in our analyses (including the examples in Table 1) are idealizations that are best treated as high-level summaries of a more complex reality. Within any individual language, there may be departures from our idealizations and these differences may be irregular (e.g., the English plural is not marked for some nouns like *deer*) or context-dependent (e.g., in Hunzib evidentiality is marked only in the past tense).

The following sections introduce additional assumptions made when analyzing each of the three grammatical features and describe our samples of attested languages. These samples are drawn from a diverse set of language families and geographic regions, but are convenience samples that aim for breadth of coverage rather than tight control over genealogical or geographic relationships. Given the nature of our analyses, controlling for historical descent and geographic region does not seem essential, and the more important question is the extent to which our samples cover the space of attested feature inventories. Some extant inventories are almost certainly missing from our samples, and it will be valuable to revisit our analyses if and when larger data sets become available.

3.1 Number

Although the underlying semantic dimension for number is probably the natural numbers,[1] we consider only natural numbers less than or equal to ten for simplicity. The universe $\mathcal{U}$ therefore includes 10 world states, one for each number considered. Number marking in English distinguishes between singular (1) and plural (> 1), but some languages have more precise systems. For example, Murrinh-Patha distinguishes between singular (1), dual (2), paucal (3-6) and plural (> 5) (Corbett, 2000). While English and Murrinh-Patha require a speaker to always use the most specific marker, some languages allow speakers a choice between specific and less specific markers. For example, Larike distinguishes between singular (1), optional dual (2), optional trial (3) and plural (> 2) which means that the plural is always an alternative to the dual and to the trial (Laidig & Laidig, 1990).

When communicating about a state including n items, the speaker distribution s_n is intended to capture the speaker's uncertainty about the precise number of items present. Speaker distributions associated with the ten possible world states are shown in the left center panel of Figure 1, and these distributions are based on data from a timed, high-contrast estimation task (Cheyette & Piantadosi, 2020). The prior distribution $p(s)$

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[1] We only consider numerical amounts in our analysis, leaving the dimension of individualization (Grimm, 2018) for future work.

captures the relative frequencies with which speakers attempt to convey the ten different meanings. Usage frequencies for number have been extensively studied using cross-linguistic corpora (Dehaene & Mehler, 1992; Piantadosi, 2016), and these studies suggest that $p(s)$ can be roughly approximated as an inverse square law (bottom left panel of Figure 1).

For our analysis, we compiled and coded the number marking inventories from 37 languages in Corbett (2000), representing 15 language families and 15 unique encoding systems. Five of these inventories are shown in Table 1. We adopt the standard linguistic conventions for number distinctions as labels (glossed in the caption). The main challenge in coding number systems is that indeterminate meanings (paucal: 3-6, plural: > 2, greater paucal: 6-8, greater plural: > 9) vary slightly across languages. For example, we code plural in Murrinh-Patha as greater than 5 because there are non-optional meanings for all lower numbers; whereas, plural in English is greater than two because there is only one other meaning. When a distinction is optional, we assume that the speaker chooses the two possible meanings equally often (see Supplemental Information for further details).

To study how meanings are realized as forms, we compiled number forms for a subset of 33 languages. Number is marked in a variety of ways across languages, and we considered only nominal and pronominal marking of grammatical number. For brevity, the systems in Table 1 show a single form for each meaning, but our data set and analyses allow for multiple forms per meaning. For example, the Russian data include number forms for different combinations of case and gender.

3.2 Tense

Tense is analogous to number but the underlying dimension is time rather than quantity. Tense marking in English distinguishes between past, present and future, but some languages have more elaborate tense inventories that specify not only whether an event is in the past or future, but also how far in the past or future it is. For example, Hixkaryana distinguishes between events in the immediate past (same day or previous night), near past (past few months) and remote past (Derbyshire, 1979). Researchers in formal semantics and artificial intelligence have developed precise representations of tense that could potentially be used in frameworks like ours (Rescher & Urquhart, 1971; McCarthy & Hayes, 1981; Allen, 1983; Reichenbach, 1947), but we take a simpler approach that can be readily applied across languages and is similar to that of Velupillai (2016b). We formulate $\mathcal{U}$ as a set of seven temporal intervals: remote past (A), near past (B) and immediate past (C), present (R), immediate future (X), near future (Y) and remote future (Z). These intervals are not sufficient to capture the tense inventory of every language in full: for example, Comrie (1985) reports that Kiksht, a language of the US Pacific Northwest, distinguishes between six or seven past tense categories. Our seven-interval timeline is therefore a pragmatic choice that allows us to represent the tense inventories of many but not all of the languages of the world.[2]

As in our number analysis, we pair each element of $\mathcal{U}$ with a speaker distribution s, and the seven meaning distributions are shown in the center panel of Figure 1. These distributions are intended to capture the uncertainty that speakers maintain over the exact time of an event: for example, s_a captures uncertainty about an event that actually took place in the remote past. To formulate these distributions we postulate major boundaries between past, present, and future, and minor boundaries between the three pasts (remote, near and immediate) and between the three futures. The distributions are defined in terms of two parameters κ and λ that specify how sharply probability mass decreases across minor and major boundaries. We set $\kappa = 0.5$ and $\lambda = 0.1$, which means that distributions drop by factors of 2 and 10 across minor and major boundaries respectively. Our results are qualitatively robust to variation in the speaker distributions as long as there is an appreciable decrease across minor boundaries ($\kappa \leq 0.75$) and a reasonable distinction between major and minor boundaries (κ and λ are not equivalent or near equivalent).

We estimated the prior $p(s)$ using a two-step process. In the first step we used estimates of past, present and future from an analysis of social media (Park et al., 2017). The resulting counts yield a distribution of $[0.274, 0.475, 0.251]$ over the coarse categories of past, present and future. Second, we used frequencies of temporal adverbs such as *yesterday*, *last week* and *last month* (see Table S1 for the complete list) to distribute probability mass among the three levels of remoteness within both past and future categories. All frequencies were derived from the Google N-gram English corpus (Michel et al., 2011) for 1985, the year of

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[2]We focus in this paper on absolute tense, leaving relative tense for future work.

publication for the source of many of our tense systems (Dahl, 1985). The prior distribution resulting from the two-step process is shown in the bottom center panel of Figure 1.

For our analysis, we compiled tense inventories for 157 languages, representing 73 language families and 16 unique inventories. Our sample was largely taken from Dahl (1985). To study how meanings are realized as forms, we compiled forms for a subset of 33 languages. Languages were selected with a bias toward languages with more linguistic forms and toward grammars that made the relevant information especially clear.

The major challenge encountered in assembling the data is that tense is often hard to separate from aspect and modality (J. L. Bybee, Perkins, & Pagliuca, 1994). For example, in some languages the primary distinction is between perfective and imperfective (roughly whether an action is complete or incomplete) rather than between past and future. Some languages include markers for categories (e.g., past perfective) that combine tense and aspect. When consulting our primary sources, we made our best judgment about whether a language could be represented in our coding scheme without distorting it too greatly, and excluded two languages (Hawaiian and Ewe) for which our scheme seemed especially inadequate. A second and less fundamental challenge is that languages which make remoteness distinctions do not include categories that are precisely equivalent. Our coding scheme distinguishes between remote (more than 7 days distant), near (between 1 and 7 days) and immediate (on the same day), and we fitted each language into this scheme as best we could.

3.3 Evidentiality

Evidentiality is a grammatical feature that conveys the source of a piece of information: for example, whether the speaker saw an event or heard it described by another person. There is no standard characterization of the space of possible sources, and we therefore formulate $\mathcal{U}$ as the set of six sources distinguished by Aikhenvald in her typology of evidential systems (Aikhenvald, 2004). In principle our framework allows these sources to be located within a multidimensional space, but for simplicity we order them along a single dimension that is consistent with Willett (1988)'s hierarchy of evidentiality values and roughly captures distance from the speaker. The first source is visual perception, and the second includes all senses other than vision. Next comes inference from visual evidence (e.g., learning there was a fire by seeing smoke), followed by assumption. The penultimate source combines general world knowledge (e.g., "it is known") and hearsay, and the final source is quoted speech. Languages group these six sources in different ways. For example, Quechua (Table 1) has markers for direct evidence (visual and sensory perception), indirect evidence (inference and assumption), and reported evidence (hearsay and quotation). In contrast, Turkish makes a simple partition between firsthand (visual sources) and non-firsthand (all other information sources).

Psychological evidence from Western populations suggests that speakers are often uncertain about the source of information retrieved from memory (Johnson, Hashtroudi, & Lindsay, 1993), but there are no detailed characterizations of how this uncertainty is distributed across different kinds of sources. We therefore specify the speaker distributions using the same hierarchical approach described for tense. Following Willett (1988)'s hierarchy, we assume major boundaries between perception (visual and sensory), reasoning (inference and assumption) and external report (hearsay and quotation speech), and minor boundaries within each of these three pairs. As for our tense analysis we use parameters κ and λ that specify how sharply probability mass decreases across minor and major boundaries. Our results are again qualitatively robust to variation in these parameters, and as before we set $\kappa = 0.5$ and $\lambda = 0.1$. The resulting speaker distributions are shown on the right center panel of Figure 1.

Although evidentiality occurs in around a quarter of the world's languages, few corpora are available for languages with fine-grained evidentiality inventories. We therefore estimate the prior $p(s)$ using a single corpus of Cuzco Quechua text (Rios, Göhring, & Volk, 2008). Quechua groups the six sources in $\mathcal{U}$ into three pairs, and we therefore divide corpus frequencies evenly within these pairs to produce the prior shown in the bottom right panel of Figure 1. For evidentiality in particular, the data available for grounding assumptions about the prior and speaker distributions are relatively limited, and our results should be viewed as tentative conclusions only.

We conducted our analysis on a set of 184 extant languages, representing 61 language families and 16 unique inventories. Descriptions of all languages were taken from Aikhenvald (2004), and five are represented in Table 1. To study how meanings are realized as forms, we compiled forms for a subset of 31 languages.
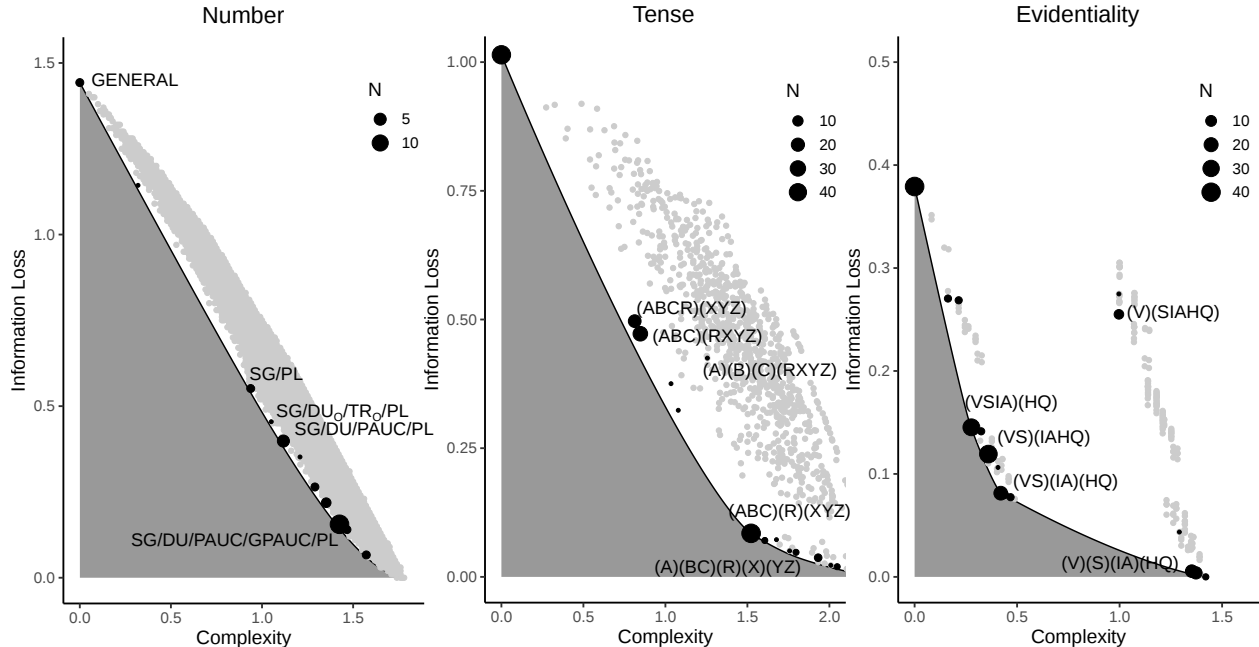
Figure 2: Analysis of the meaning of grammatical markers. (a)-(c) Trade-offs between information loss and complexity for number, tense and evidentiality. Attested inventories (black points) and unattested systems (grey points) are plotted in the space of all possible grammatical systems. Systems that achieve optimal trade-offs lie along the Pareto frontier (solid line), and the shaded region below the line shows trade-offs that are impossible to achieve.

Similar to tense, there was some difficulty separating evidentiality from other grammatical features including mood, mirativity (grammaticalized surprise), aspect and tense. For example, in Mansi, Svan and Turkish, evidentiality correlates with mirativity, and the non-firsthand marker can be used when the speaker has perfect visual evidence but witnesses something so surprising that they do not believe it. Evidentiality can also interact with genre, register and person systems. For example, in Meithei, the source of information is not always with respect to the speaker (first person) but can be calculated with respect to the listener (second person). As with tense, we tried our best to encode each system as faithfully as possible, but acknowledge that our encoding of evidentiality represents a starting point only and that future work will be required (see Supplemental Information for further discussion).

4   Analysis of meaning

We first analyze the feature values or meanings captured by each language in our data set, and the next section analyzes the forms that realize these meanings. For each of the three grammatical features, the space of possible encoders q_m is shown in Figure 2. Encoders that achieve optimal trade-offs between information loss and complexity lie along the Pareto frontier, shown here as a solid line, and the dark grey region below the curve shows trade-offs that are impossible to achieve. Attested inventories are shown as black points, and the light gray points include all possible inventories that partition $\mathcal{U}$ into non-overlapping feature values. Attested inventories (black points) are generally closer to the Pareto frontier than are unattested inventories (light gray points), suggesting that attested inventories for number, tense and evidentiality are near-optimal. The Supplemental Information includes a quantitative analysis that supports this conclusion strongly for number and tense and less strongly for evidentiality. It also shows that our model accounts better for attested inventories than an alternative approach previously applied to tense marking (Bacon, 2020) that defines the complexity of an inventory as the number of markers that it includes.

Although most attested inventories lie close to the Pareto frontier, there are a handful of notable exceptions. There are no clear outliers for number, and for tense, the single outlier is Hixkaryana. The Hixkaryana tense inventory (see Table 1) is unusual because it includes a relatively large number of categories but does
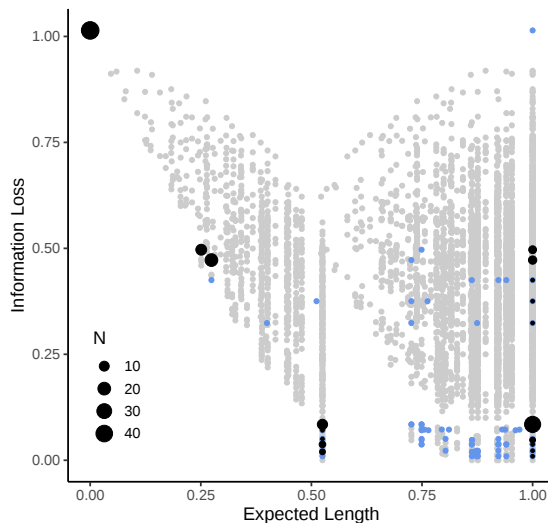
Figure 3: Zero marking analysis of tense systems. Trade-off between information loss and expected length when zero-marking is allowed for all tense systems ($N = 157$). Black dots show attested systems (size denotes frequency), blue dots show all ways to zero-mark at most one feature value in an attested system, and grey dots show possible but unattested systems.

not distinguish between present and future. For evidentiality, there are two notable groups of outliers driven by the same principle: our model predicts that distinctions at the level of Willett (1988)'s clusters should be made before distinctions within these clusters. For a few two term systems, including Turkish, Mansi and Meithei, and one three term system, Siona, a distinction between visual and sensory information is made before distinctions are made between all of Willett (1988)'s levels. For a more detailed discussion of individual languages and outliers see the Supplemental Information. The general conclusion from this discussion is that most attested systems are qualitatively similar to optimal systems.

For each of the plots in Figure 2, traversing the Pareto frontier from top left to bottom right generates a hypothetical evolutionary trajectory that makes predictions about the order in which distinctions are introduced if a system grows more complex over time (Zaslavsky et al., 2018; Zaslavsky, 2020). The Supplemental Information includes a detailed analysis of these trajectories and shows that they recapitulate some patterns previously identified by work in linguistic typology. Following Greenberg (1963), these patterns are often formulated as universal constraints on possible systems: for example, one such universal states that if a number system has a trial, then it also has a dual. Our theory broadly captures this and other known patterns, but is most compatible with the view that they are strong regularities that emerge from soft functional constraints instead of strict universals that hold without exception (M. S. Dryer, 1998; N. Evans & Levinson, 2009).

5 Analysis of form

We now analyze form length for number, tense and evidentiality. Before comparing form lengths across languages, we normalize lengths within each system to allow for the fact that lengths may be systematically longer in some languages (e.g., those with relatively small phoneme inventories) than others. Our first analysis asks whether feature values that are "zero-marked" (i.e. not overtly expressed, as for the nominal singular marker in English) tend to be more frequent than other feature values belonging to the same system (Greenberg, 1966; Haspelmath, 2021). To address this question we use a coarse form of normalization that assigns a length of 0 to any feature value that is zero-marked and lengths of 1 to all other feature values. Figure 3 plots the information loss from our analysis of meaning against expected length for tense, and therefore shows how informativeness of meaning trades off against brevity of form. The Supplemental Information contains an analogous plot for evidentiality but not number, because our sample of number markers includes
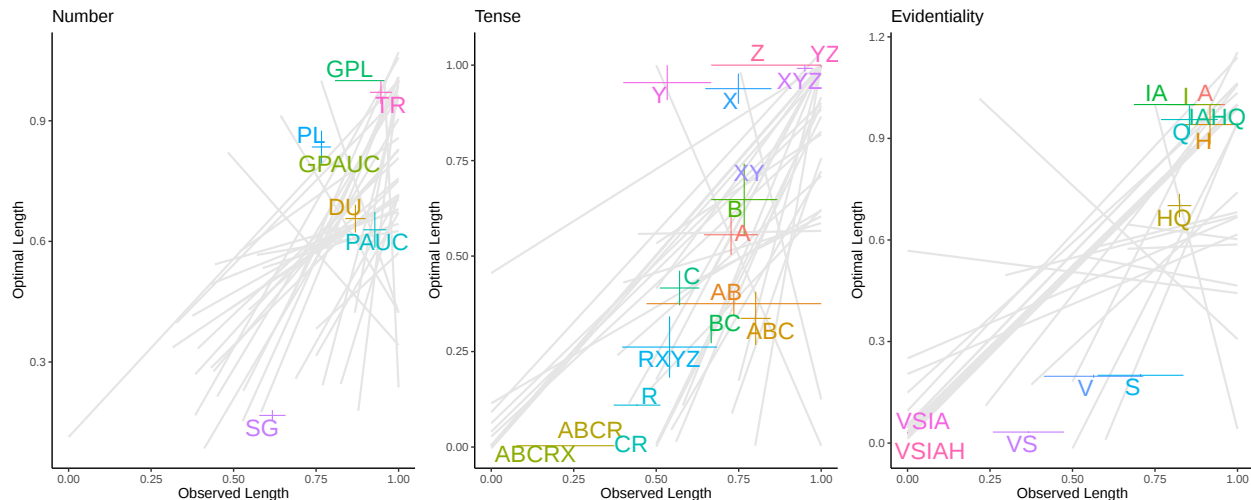
Figure 4: Analyses of the form of grammatical markers. Relationship between optimal and observed codelengths for a subsample of number, tense and evidentiality systems. Within each language, forms were unit normalized and the lengths of multiple forms for the same feature value within a language were averaged. Gray lines show trend lines for each language, and each colored data point shows an average across all languages that include a given feature value. Error bars show standard error of the mean, and the vertical error bars occur due to normalization and because the optimal length for a marker (e.g., past, or abc) depends on whether or not it is optional. Colors are arbitrary but help to distinguish overlapping error bars.

pronominal forms for which zero-marking does not occur. The black dots represent attested systems, and the light blue dots include all permutations of systems that use zero marking for at most one category in an attested system. The small grey dots show all ways to apply zero-marking to unattested systems. Attested systems with zero marking overwhelmingly tend to zero-mark the most frequent feature value and therefore lie along the Pareto frontier. The remaining attested systems explicitly mark all grammatical feature values and appear as a column of black dots with expected length equal to one. The Supplemental Information includes a statistical analysis suggesting that whether or not a tense system uses zero-marking can be partially predicted by the information loss of the meaning encoding system. When information loss is high, zero-marking provides relatively large reductions in expected length and is relatively likely to be used. In contrast, systems with low information loss have little to gain by zero-marking and are relatively likely to be explicit.

We now ask more generally whether the frequency of a grammatical marker is inversely related to the length of its form. Form length should ideally be measured in phonemes, but we do not have phonemic transcriptions for all languages in our samples and therefore use orthographic length as a rough proxy for phonemic length. Within each system, form lengths are normalized so that the longest form has length 1. We compare these "observed lengths" to predicted or "optimal lengths," where the optimal length for a marker with probability $p(m)$ is the surprisal $-\log(p(m))$. Frequent markers have short optimal lengths, and the optimal length of each marker can be interpreted as the number of bits used to represent the marker given an optimal code. Figure 4 shows that observed and optimal lengths are positively correlated across our samples of number, tense and evidentiality systems, and correlations for selected languages are shown in the final column of Table 1. The labeled data points in Figure 4 are based on averages across all systems that share a given feature value, and the gray lines are regression lines based on lengths from individual languages. Some individual languages (gray lines) represent exceptions to the general trend — for example, Tamil and Seneca have slightly shorter forms for future tense than for past (ABC) or past/present (ABCR) tenses. In general, however, languages tend to assign relatively short forms to markers that are high in frequency.

The results in Figure 4 are highly compatible with previous discussions of coding efficiency and grammatical marking. Most relevant to our approach is the work of Haspelmath (2021), who uses a broad range of grammatical patterns to demonstrate that more frequent grammatical feature values tend to have relatively short forms, and explains this result using the same functional-adaptive principles invoked by our theory.

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Relative to this prior work our main contribution is to suggest that coding efficiency should not be considered in isolation, but rather trades off against a pressure for informative communication.

6   Discussion

We presented an account of grammatical marking which suggests that number, tense and evidentiality systems across languages achieve efficient tradeoffs between informativeness and simplicity. Our results align with related results previously reported for domains including color naming (Zaslavsky et al., 2018), kin naming (Kemp & Regier, 2012), quantifiers (Steinert-Threlkeld, 2020), numeral systems (Xu, Liu, & Regier, 2020), and indefinite pronouns (Denić, Steinert-Threlkeld, & Szymanik, 2020), and with a broader literature that characterizes ways in which language supports efficient communication (Gibson et al., 2019). Within this literature there are studies that focus on meaning (e.g., Kemp et al., 2018) and studies that focus on form (e.g., Piantadosi et al., 2011), but few that address both meaning and form (e.g., Dautriche, Mahowald, Gibson, & Piantadosi, 2017; Tamariz, 2008). Our work suggests how form and meaning can be brought together in an integrated information-theoretic framework.

Our analysis assumed that the function of each grammatical feature is to convey information about a single underlying dimension. Our approach, however, can be directly applied to settings in which the conceptual universe $\mathcal{U}$ combines multiple semantic dimensions: for example, both person and number (Zaslavsky, Maldonado, & Culbertson, 2021). Applying the framework in this way provides a new perspective on grammatical paradigms, and may help to explain attested patterns of syncretism. A further possible extension is to allow for additional functions that grammatical features may serve: for example, some features (e.g., case; Mollica & Kemp, 2020) may convey information about structural dependencies via indexing, and others (e.g., grammatical gender; Dye, Milin, Futrell, & Ramscar, 2017) may convey information about what forms should be expected next. Frequent linguistic units such as grammatical markers are especially likely to have multiple functions (Haspelmath, 2003; Zipf, 1949), and capturing the full range of these functions is a major challenge for quantitative approaches.

Our analysis focused only on grammatical marking but other linguistic strategies are available for communicating about number, time, and information source, including the use of quantifiers, temporal adverbs, and modal verbs. The languages in our data sets that do not mark number, tense and or evidentiality rely on these other strategies. Studying grammatical marking and other individual strategies in isolation is a natural first step, but future work should aim to allow for multiple different strategies when evaluating communicative efficiency.

Our work suggests that systems of grammatical markers achieve efficient trade-offs between informativeness and simplicity, but does not capture the historical processes that led to this outcome. It is possible that efficient trade-offs could arise in the absence of communicative pressures (Caplan, Kodner, & Yang, 2020), but recent work on cultural evolution and language acquisition suggests that language learning and use impose pressures towards informativeness and simplicity (Kirby, Tamariz, Cornish, & Smith, 2015; Carstensen, Xu, Smith, & Regier, 2015; Carr, Smith, Culbertson, & Kirby, 2020). On this account, the pressure towards informativeness applies during cooperative language use (e.g., Fay, Garrod, Roberts, & Swoboda, 2010), and the pressure towards simplicity applies during language learning (e.g., Hudson Kam & Newport, 2005). There is now a sizable body of evidence in language acquisition showing that learners reshape their input to learn languages that are simpler, easier to produce and more informative than their input (Fedzechkina et al., 2012; Kanwal et al., 2017; Kurumada & Grimm, 2019; Fedzechkina & Jaeger, 2020; Culbertson & Smolensky, 2012).

Connecting our approach with a model of historical language change may help to address two additional questions left open by our results. Our approach helps to explain the range of grammatical systems observed across languages, but does not explain why some systems are more frequent than others, or why any particular language has the systems that it does. One possibility is that different cultures impose different functional constraints, but a second possibility is that variation across languages reflects a set of crystallized historical accidents. If grammatical systems were initialized randomly, selective pressures over time may lead them to converge on a relatively small set of attractors, and the relative frequencies of these attractors could be explained by the relative sizes of their basins of attraction. Phylogenetic analyses have provided insight into the evolution of both semantic and grammatical systems (Jordan, 2011; Haynie & Bowern, 2016; M. Dunn,

Greenhill, Levinson, & Gray, 2011), and a similar approach could be productively applied to our data.

Although we focused on number, tense and evidentiality marking, our general approach can be applied both to other grammatical features and to the lexicon. In all cases, the goal is to simultaneously explain both the meanings captured by a linguistic system and the relative lengths of the forms that express those meanings. Grammar and the lexicon have traditionally been explored somewhat separately, but information-theoretic analyses can help to characterize how both support efficient communication.

7 Materials & Methods

7.1 Treatment of Data

Languages were primarily sampled from monographs surveying the grammatical features of number (Corbett, 2000), tense (Dahl, 1985), and evidentiality (Aikhenvald, 2004) with the goal of including as many distinct attested systems as possible. All data and code used in the analyses are available in an OSF repository: https://osf.io/s5b7h/

7.2 Specifying encoder distributions

Because usage data are not available for many languages in the data set, encoders $q(m|s)$ were determined using a maximum entropy assumption. This assumption is only relevant for languages that have optional distinctions. The unique encoders in our analysis are shown in detail in the Supplemental Information.

7.3 Speaker Distributions

For number, the nth speaker distribution is given by the analytical form:

$$S(u|n) \propto p(u) \exp\left(-\frac{p(n)}{\lambda_n}(n-u)^2\right), \tag{4}$$

where the λ_i are empirically estimated precision parameters. To avoid having speaker distributions with no uncertainty due to values smaller than numerical precision, we added 10^{-5} to each state u and renormalized. For tense and evidentiality, the speaker distribution associated with state u^* is given by:

$$S(u|u^*) \propto B^\lambda b^\kappa, \tag{5}$$

where B and b are the number of major and minor boundaries separating u and u^* and λ and κ are the discount rates across major and minor boundaries respectively. For our analyses, λ and κ were set to 0.1 and 0.5 respectively.

7.4 The IB Pareto-frontier

The trade-off between complexity and information loss is given by the IB objective function (Tishby et al., 1999):

$$\mathcal{F}_\beta[q(m|s)] = I(M;S) - \beta I(M;U). \tag{6}$$

Following Zaslavsky et al. (2018), the Pareto-frontier was computed using reverse deterministic annealing (Tishby et al., 1999). For number, the β schedule was 2^x for x from 4 to 0 by 0.001 increments. For tense, the β schedule was 2^x for x from 5 to 0 by 0.001 increments. For evidentiality, the β schedule was 2^x for x from 5 to 0 by 0.001 increments.

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Supplemental Information

S1   Representing grammatical systems

As described in the main text, for each grammatical feature we define an underlying semantic dimension that specifies possible states of the world. For each attested system, we define the language's encoder $q_m(m|s)$ using a maximum entropy assumption over the meanings m that can be used to convey a speaker distribution s. For example, in the Sanskrit number system two objects can only be communicated using the dual so $q(\text{DUAL}|s_2) = 1$. In Slovene, however, the dual is optional and so s_2 can be expressed using either the dual or the plural. In this case, the maximum entropy assumption dictates that $q(\text{DUAL}|s_2) = q(\text{PL}|s_2) = 0.5$. Section S3.2 includes an analysis suggesting that our conclusions are robust to variations in the maximum entropy assumption.

When enumerating hypothetical systems (grey dots in Figure 2) we include all systems that partition the underlying semantic dimension. Systems with optional markers could be considered in principle, but doing so would make the analysis intractable. In general, allowing the same state to be labelled with two or more markers decreases both complexity and communicative precision. As a more stringent test of our theory, section S3.3 compares attested systems against permutations of attested systems, and these permutations represent a subset of all possible optional partitions. Further, as attested languages often partition the semantic dimension into connected regions, section S3.4 repeats our comparisons of attested systems against all possible partitions and against permutations of attested systems but imposes an additional semantic connectivity constraint.

As described in the main text, temporal adverbs used to estimate the prior $P(s)$ for tense are listed in Table S1.

Deg. of remoteness	Temporal adverbs
immediate	*today*
near past	*yesterday*
near future	*tomorrow*
remote past	*last week/month/year/decade/century*
remote future	*next week/month/year/decade/century*

Table S1: Temporal adverbs used to estimate usage probabilities for locations along the timeline. Because the immediate past and immediate future are both expressed through *today* in English, we assigned half of *today*'s frequency to each of the two temporal locations.

We coded the forms of grammatical markers f in order to explore the relationship between empirical and optimal codelengths. Grammatical features are often realized within a paradigm, which can result in multiple forms for the same grammatical feature value (see Table S2). For our analyses, the observed length of a feature value is defined as an average over the lengths in characters for all forms of that feature value. The observed lengths in Figure 4 are normalized for each language by dividing by the maximum observed length for that language.

We gathered forms for grammatical features using monographs on each grammatical feature as a starting point (number: Corbett, 2000, tense: Dahl, 1985, evidentiality Aikhenvald, 2004). If the monograph contained forms for a system, we used those forms. Additionally, we sought out source grammars, where possible, for each language to extract forms from a primary source. To allow a strong test of our theory we focused on languages with a relatively large number of forms. Tense and evidentiality are predominantly realized as grammatical morphemes that attach to the verb. The realization of grammatical number is more varied across languages. Number can be marked on nouns, pronouns, verbs, adjectives, determiners, case markers and more. Further, number marking can be lexicalized or realized via additive morphology (affixes), reduplication, or suppletion (for a survey of grammatical pluralization strategies see Chen, 2019). We therefore compiled a list of all possible ways that number is marked in our sample of languages (available on OSF). Based on this survey, we chose to focus on nominal and pronominal marking in order to cover as many languages as possible from our sample (33 out of 37). In doing so our approach differs from previous studies

that have focused either on the marking of specific feature values (e.g., plurality Chen, 2019) or on the notion of "markedness" (Brunner, 2010; Haspelmath, 2021). Nonetheless, we arrive at similar conclusions.

Future		First	Second	Third
	SG	*habl**aré***	*habl**arás***	*habl**ará***
	PL	*habl**aremos***	*habl**aréis***	*habl**arán***
Present	SG	*habl**o***	*habl**as***	*habl**a***
	PL	*habl**amos***	*habl**áis***	*habl**an***
Past Perfect	SG	*habl**é***	*habl**aste***	*habl**ó***
	PL	*habl**amos***	*habl**astais***	*habl**aron***
Past Imperfect	SG	*habl**aba***	*habl**abas***	*habl**aba***
	PL	*habl**ábamos***	*habl**abais***	*habl**aban***

Table S2: Example paradigm for for tense forms in Spanish.

S2   Alternate Measures of Complexity
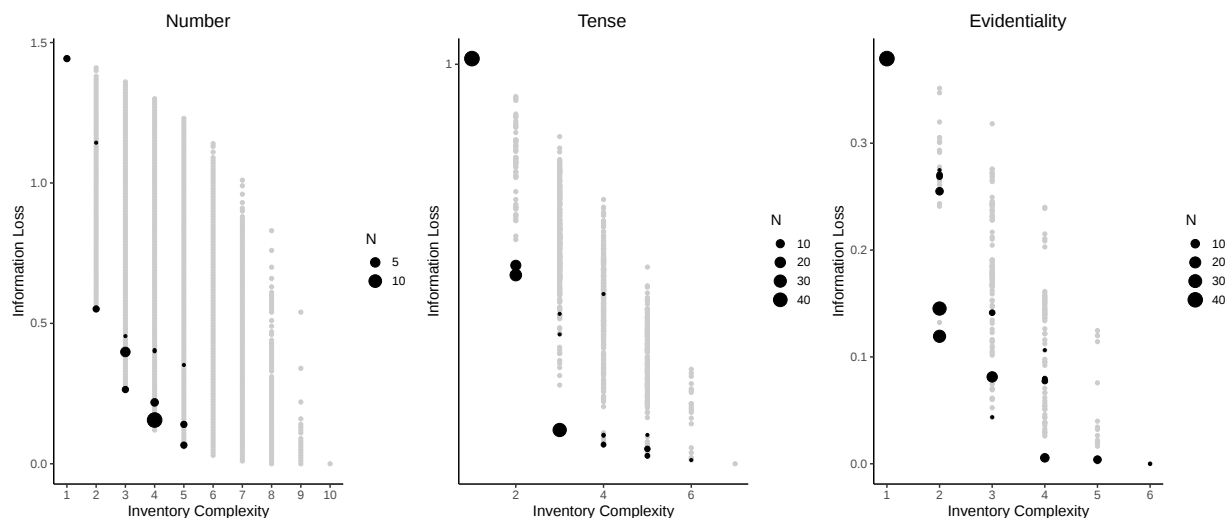
Figure S1: Trade-off between information loss and inventory complexity.

Our metric of complexity is based on an information-theoretic analysis of communication, but linguists have proposed many alternative metrics of morphological complexity (e.g., Baerman, Brown, & Corbett, 2015; Stump, 2017). Nichols (2019) discusses two general ways of measuring complexity: inventory complexity, or the number of unique morphemes in a system, and descriptive complexity (Kolmogorov Complexity), or the minimal amount of information required to describe a system. Figure S1 shows the tradeoff between information loss and complexity when complexity is defined as inventory complexity, and the same approach was previously applied to tense marking in Bacon (2020). The results still provide some evidence that attested systems of a given complexity level tend to minimize information loss, which is not surprising as measures of morphological complexity tend to be correlated (Bentz, Soldatova, Koplenig, & Samardžić, 2016). The results, however, reveal that replacing our information-theoretic complexity measure (Equation 1) with inventory complexity means that departures from the Pareto frontier become more common and more substantial, especially for systems with low inventory complexity. The results therefore suggest that our model accounts better for attested systems than a similar approach that relies on inventory complexity.

Some previous work that explores tradeoffs between information loss and complexity has relied on descriptive complexity (Kemp & Regier, 2012; Denić et al., 2020; Steinert-Threlkeld, 2020), and a similar

approach could be applied here. Previous work in this tradition typically commits to a domain-specific representation language, and complexity is then defined as the length of the shortest representation in that language. Extending this approach to our setting would probably require three separate representation languages for number, tense and evidentiality. Formulating these representation languages seems relatively challenging, and formulating the domain-specific components of our approach (the speaker distributions s) seems straightforward by comparison.

An alternative approach is to invoke a universal representation language with the property that the length of an object's representation is inversely proportional to its probability. Approaching the problem in this way allows theoretical connections to be established between descriptive complexity and information-theoretic measures of complexity (Grunwald & Vitányi, 2004). From this perspective, our complexity measure defined in Equation 1 can be viewed as a kind of descriptive complexity measure.

S3   Quantitative Analyses

To further explore the results in Figures 2, 3 and 4, we ran a series of quantitative evaluations.

S3.1   Evaluation of Near-Optimality
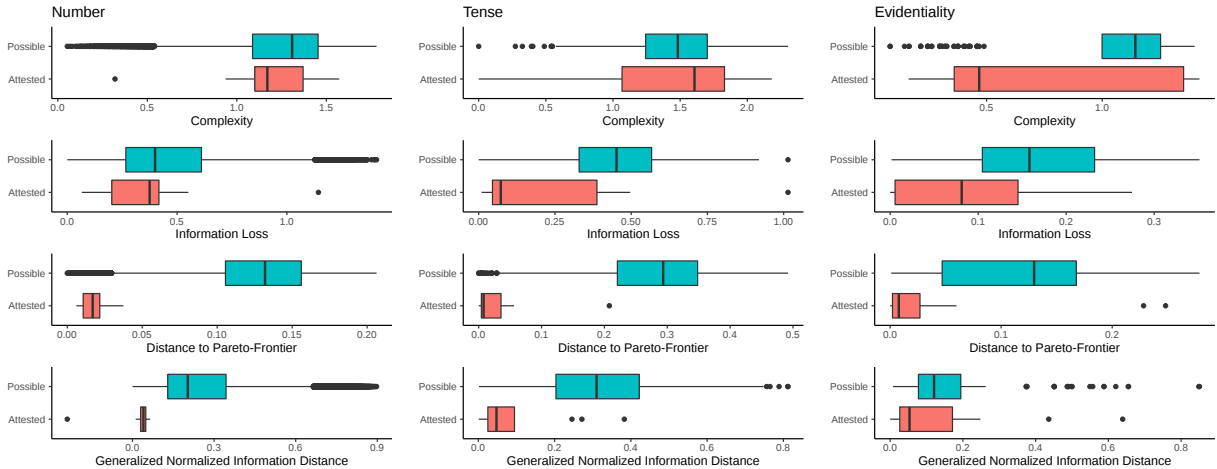
Figure S2: The distributions of different evaluation metrics for unattested and attested grammatical systems.

Figure 2 suggests that attested systems achieve near-optimal tradeoffs between the dimensions of information-loss and complexity. A tradeoff implies that both dimensions matter, and the signature of this tradeoff is that attested systems tend to lie near the Pareto frontier. The penultimate column in Table 1 shows the Euclidean distance between a system and our 2-dimensional Pareto frontier, but here we also consider the generalized Normalized Information distance (gNID), an alternative measure introduced in Zaslavsky et al. (2018). Considering gNID allows us to directly quantify the similarity between attested systems and their closest optimal counterparts. For completeness, we briefly present the formal definition of gNID (see the SI of Zaslavsky et al., 2018 for more detail). Let m_1 and m_2 be markers drawn from encoders q_1 and q_2 respectively, where

$$q(m_1, m_2) = \sum_s p(s) q_1(m_1|s) q_2(m_2|s). \tag{7}$$

The distance between q_1 and q_2 is low if the mutual information $I(M_1; M_2)$ is high, and the gNID is calculated using

$$\text{gNID}(q_1, q_2) = 1 - \frac{I(M_1; M_2)}{\max\{I(M_1; M_1'), I(M_2; M_2')\}}. \tag{8}$$

If attested system q resembles a system q^* on the Pareto frontier, then gNID(q, q^*) will be low. We calculate gNID for each attested system q with respect to the optimal system q^* that minimizes gNID.

Figure S2 compares attested and unattested inventories with respect to four evaluation metrics: complexity, information loss (KL Divergence), distance to the Pareto frontier, and generalized normalized information distance (gNID). Comparing these evaluation metrics allows us to see whether attested inventories are near-optimal with respect to one dimension alone, or whether attested inventories navigate a tradeoff between information loss and complexity.

	Number	Tense	Evidentiality
Complexity	-7632 (15.3)	-61.5 (0.040)	-12.6 (0.128)
Information Loss	-7986 (10.1)	-48.5 (0.313)	-13.9 (0.070)
Distance to Pareto Frontier	**-547 (7.02)**	**-16.9 (0.500)**	**-12.0 (0.233)**
gNID	-3162 (42.6)	-39.5 (0.453)	-14.5 (0.064)

Table S3: Model comparison using log likelihood values averaged across CV runs. Higher values are considered better fit. Standard errors in parentheses.

To quantitatively test for near optimality, we fit four logistic regression models using each metric to predict whether or not a system is attested. To adjust for the imbalance in number of attested and possible systems, we over-sampled the attested systems (Menardi & Torelli, 2014) and compared models using the average log likelihood from 10 runs of 5-fold cross-validation, repeated 10 times. All models were implemented in R (R Core Team, 2020) using the `tidymodels` (Kuhn & Wickham, 2020), `themis` (Hvitfeldt, 2020) and `lme4` (Bates, Mächler, Bolker, & Walker, 2015) packages. As can be seen in Table S3, Euclidean distance to the Pareto frontier better predicts the diversity of attested grammatical systems than either dimension of the trade-off alone.

S3.2 Relaxing the maximum entropy assumption for optional distinctions

		Number	Tense
Favored large	Complexity	-7591 (15.3)	-59.7 (0.035)
	Information Loss	-8021 (6.61)	-50.7 (0.31)
	Distance to Pareto Frontier	**-339 (5.44)**	**-17.1 (0.55)**
Disfavored large	Complexity	-7783 (13.2)	-59.6 (0.04)
	Information Loss	-8027 (6.67)	-50.9 (0.30)
	Distance to Pareto Frontier	**-674 (9.43)**	**-17 (0.55)**

Table S4: Model comparison with weighted optional encoding that either favors ($\approx 75\%$) or disfavors ($\approx 25\%$) the larger partition. Values are log likelihoods averaged across CV runs. Higher values are considered better fit. Standard errors in parentheses.

To assess the sensitivity of our conclusions to our max entropy assumption, we replicated our near optimality analysis assigning weights that either favor the larger partition ($\approx 75\%$) or disfavor the larger partition ($\approx 25\%$). Regardless of the weighting, Table S4 shows that Euclidean distance to the Pareto frontier better predicts the diversity of attested grammatical systems than either dimension of the trade-off alone.

S3.3 Permutation Analysis for Near-Optimality

As mentioned above, our hypothetical systems included all possible partitions of the semantic space. Some of these systems may be linguistically implausible, including the number system with ten different meanings associated with each of the ten world states. We therefore conducted a second analysis using a more restricted set of comparison systems that includes only permutations of attested systems (Zaslavsky, Regier, Tishby, & Kemp, 2019). Each permuted system is created by starting with an attested system and then permuting the underlying semantic space. Figure S3 is similar to Figure 2 but shows all possible permutations in
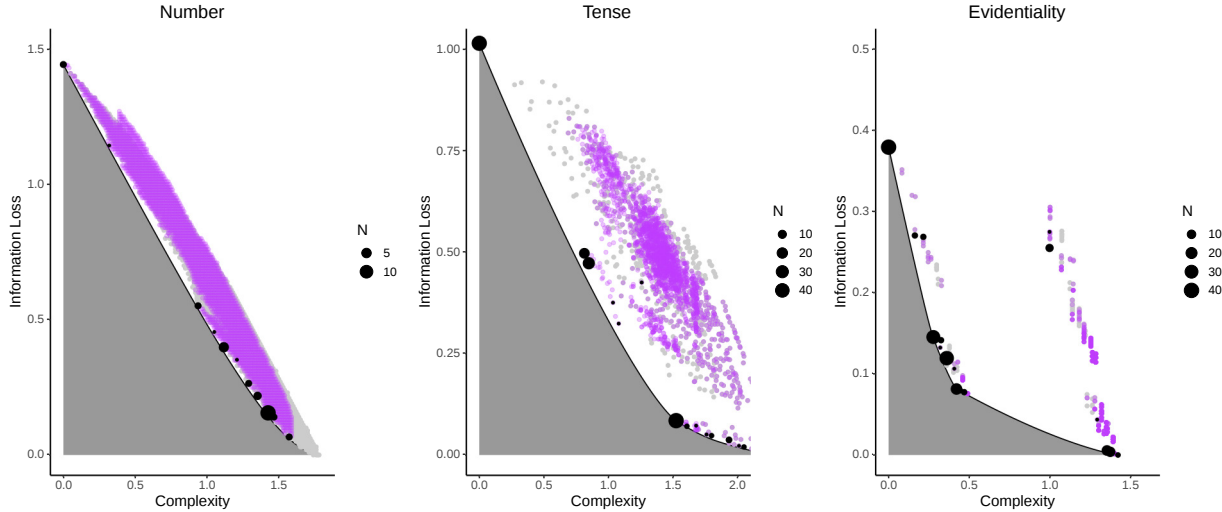
Figure S3: Permutation analysis of the meaning of grammatical markers. Top panels: (a)-(c) Trade-offs between information loss and complexity for number, tense and evidentiality. Attested inventories (black points), permuted systems (purple points) and unattested systems (grey points) are plotted in the space of all possible grammatical systems. Systems that achieve optimal trade-offs lie along the Pareto frontier (solid line), and the shaded region below the line shows trade-offs that are impossible to achieve.

purple, and Figure S4 compares attested and permuted inventories with respect to our evaluation metrics. We leave out the gNID predictor for number, as calculating gNID for all possible permutations would be computationally expensive.

	Number	Tense	Evidentiality
Complexity	-19484 (27.1)	-159 (0.195)	**-39.8 (0.331)**
Information Loss	-17567 (58.8)	-114 (0.786)	-50.8 (0.125)
Distance to Pareto Frontier	**-1124 (11.5)**	**-36.9 (0.840)**	-49.3 (0.184)
gNID	NA	-81 (0.912)	-48.5 (0.156)

Table S5: Permutation model comparison using log likelihood values averaged across CV runs. Higher values are considered better fit. Standard errors in parentheses.

Again, to quantitatively test for near optimality, we fit four logistic regression models using each metric to predict whether or not a system is attested. To adjust for the imbalance in number of attested and permuted systems, we over-sampled the attested systems (Menardi & Torelli, 2014) and compared models using the average log likelihood from 10 runs of 5-fold cross-validation, repeated 10 times. All models were implemented in R (R Core Team, 2020) using the `tidymodels` (Kuhn & Wickham, 2020), `themis` (Hvitfeldt, 2020) and `lme4` (Bates et al., 2015) packages. As can be seen in Table S5, Euclidean distance to the Pareto frontier better predicts the diversity of attested grammatical systems than either dimension of the trade-off alone for number and tense. For evidentiality, complexity alone emerges as the best predictor.

S3.4 Adding a semantic connectivity constraint

We can call a system *connected* if it partitions a semantic space into a set of connected regions. The attested languages in our tense and number samples are all connected, but the analyses in Tables S3 and S5 both compare attested systems against alternatives that mostly do not satisfy this constraint. To enable stronger tests of our theory we therefore repeated these analyses but filtered the comparison set (either all possible partitions or permutations of attested systems) to retain only connected systems. As seen in Table S6, for number and tense Euclidean distance to the Pareto-frontier was a better predictor of attested systems than
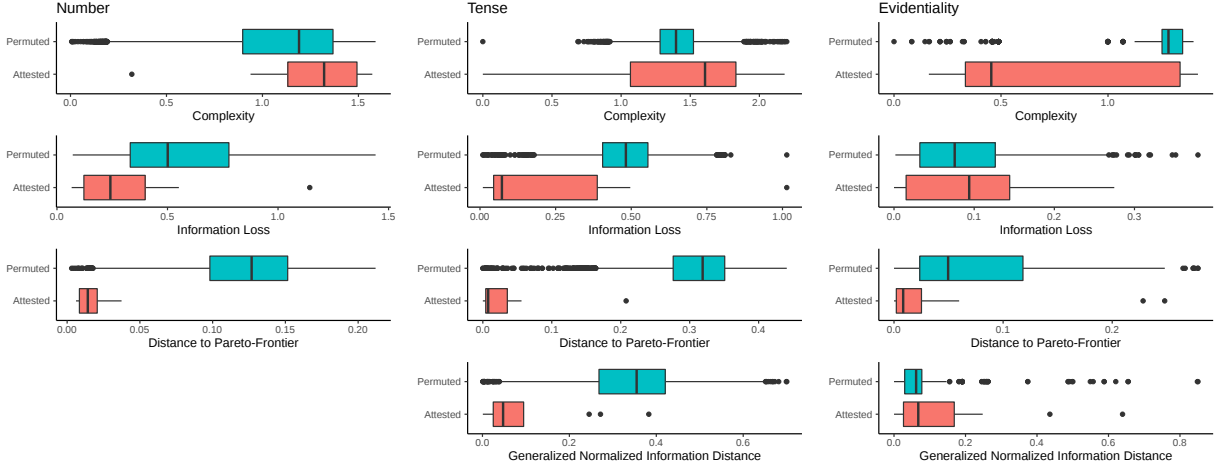
Figure S4: The distributions of different evaluation metrics for permuted and attested grammatical systems.

		Number	Tense	Evidentiality
Possible Systems	Complexity	-38.6 (0.07)	-5.81 (0.05)	-2.87 (0.13)
	Information Loss	-37.2 (0.07)	-5.01 (0.08)	**-2.58 (0.05)**
	Distance to Pareto Frontier	**-22.9 (0.47)**	**-4.30 (0.23)**	-3.18 (0.13)
	gNID	NA	-4.47 (0.19)	-2.88 (0.16)
Permuted Systems	Complexity	-42.6 (0.32)	-8.75 (0.05)	-9.92 (0.16)
	Information Loss	-45.0 (0.33)	-10.2 (0.07)	**-9.3 (0.07)**
	Distance to Pareto Frontier	**-30.0 (0.44)**	**-7.85 (0.22)**	-11.1 (0.05)
	gNID	NA	-7.92 (0.11)	-11.8 (0.06)

Table S6: Model comparison with a semantic connectivity constraint. Values are log likelihoods averaged across CV runs. Higher values are considered better fit. Standard errors in parentheses.

either dimension alone. For evidentiality, information loss was the best predictor for distinguishing between possible and permuted systems. This result therefore suggests that evidentiality systems may primarily be driven by minimizing information loss under the constraint that feature values correspond to connected regions of the semantic space. The permutation analysis without the connectivity constraint indicated that complexity was the best predictor of evidentiality systems (Table S5), but adding the connectivity constraint reduces variation in complexity and therefore yields a different conclusion.

S3.5 Evaluation of Tradeoff between Zero Marking and Information Loss

To test whether information loss influences the pattern of zero-marking in tense systems, we ran a logistic mixed effect regression model predicting zero-marking (present/absent) according to Dahl (1985) as function of information loss with a random intercept for language family. The model was implemented in R (R Core Team, 2020) using the `lme4` package (Bates et al., 2015). For all `lme4` model comparisons, p-values are calculated from a likelihood ratio test between models with and without the predictor of interest. Languages with maximal simplicity (i.e. with only one feature value) were removed for this analysis. Languages are more likely to use zero-marking at high information loss than at low information loss ($\beta = 10, \mathrm{SE} = 2.3, p < 0.05$). Because our sample was not chosen to reflect typological frequency of all attested tense systems, this finding should be treated as preliminary.

Figure S5 is analogous to Figure 3 but is based on the subset of our evidentiality dataset for which we have forms. This subset includes 15 languages with zero-marking and 16 languages without zero marking. A language was classified as zero-marked if there was at least one uninflected grammatical feature value. As for tense, we find that languages are more likely to use zero-marking at high information loss than at low
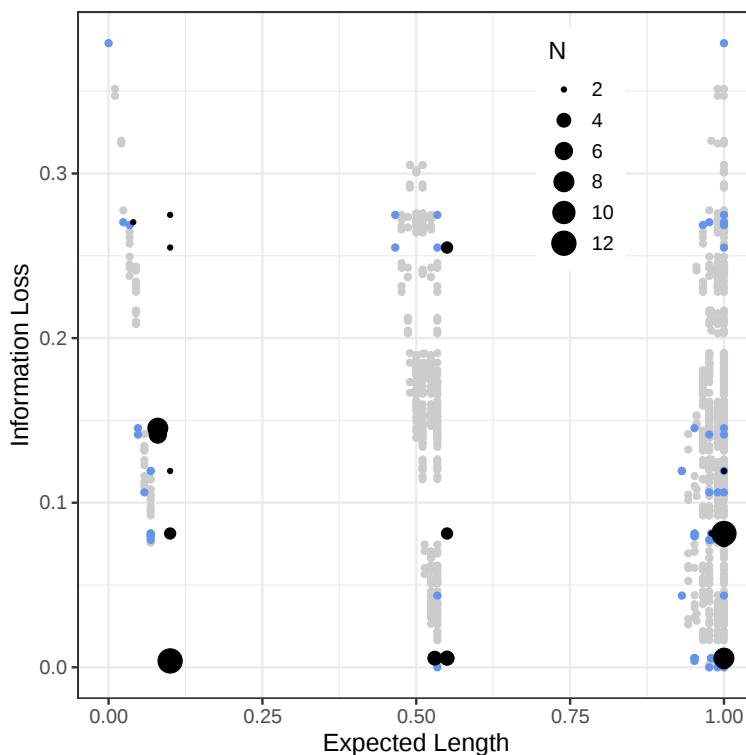
Figure S5: Tradeoff between expected length and information loss for evidentiality. The size of the black dots reflect number of attested systems. The blue dots show all ways to zero-mark at most one feature value in an attested system. The grey dots show all ways to apply zero-marking to unattested systems. The column of attested systems with expected length equal to 1 includes systems that do not use zero marking.

information loss ($\beta = 15.6, \text{SE} = 4.6, p < 0.05$) Unlike tense, however, we specifically selected the subset of evidentiality languages to favor languages with explicit forms. The results of the regression analysis for evidentiality should therefore be interpreted with caution.

Because we included both nominal and pronominal forms for grammatical number, our sample of number systems did not include zero marking. Previous work, however, suggests that the singular is sometimes zero marked but that zero marking of other feature values is extremely rare (Greenberg, 1963). A standard explanation for this finding is that zero marking tends to be applied to the most frequent feature value (here the singular; Greenberg, 1966; Haspelmath & Karjus, 2017), and the same explanation follows from our theoretical framework.

S3.6   Evaluation of Correlation between Forms and Optimal Codelengths

Figure 4 compares optimal and observed form lengths across all languages for which we compiled forms, and results for individual languages are shown in Figures S6 - S8.

To test for a relationship between observed and optimal codelengths, we conducted separate linear mixed effect regression models for each grammatical feature predicting observed length for each feature value as a function of optimal codelength with random intercepts for Language and Language Family. The models were implemented in R (R Core Team, 2020) using the `lme4` package (Bates et al., 2015). We find a significant linear relationship between optimal codelengths and observed form lengths for number($\beta = 0.29, \text{SE} = 0.05, p < 0.05$), tense ($\beta = 0.55, \text{SE} = 0.08, p < 0.05$) and evidentiality ($\beta = 0.47, \text{SE} = 0.06, p < 0.05$).
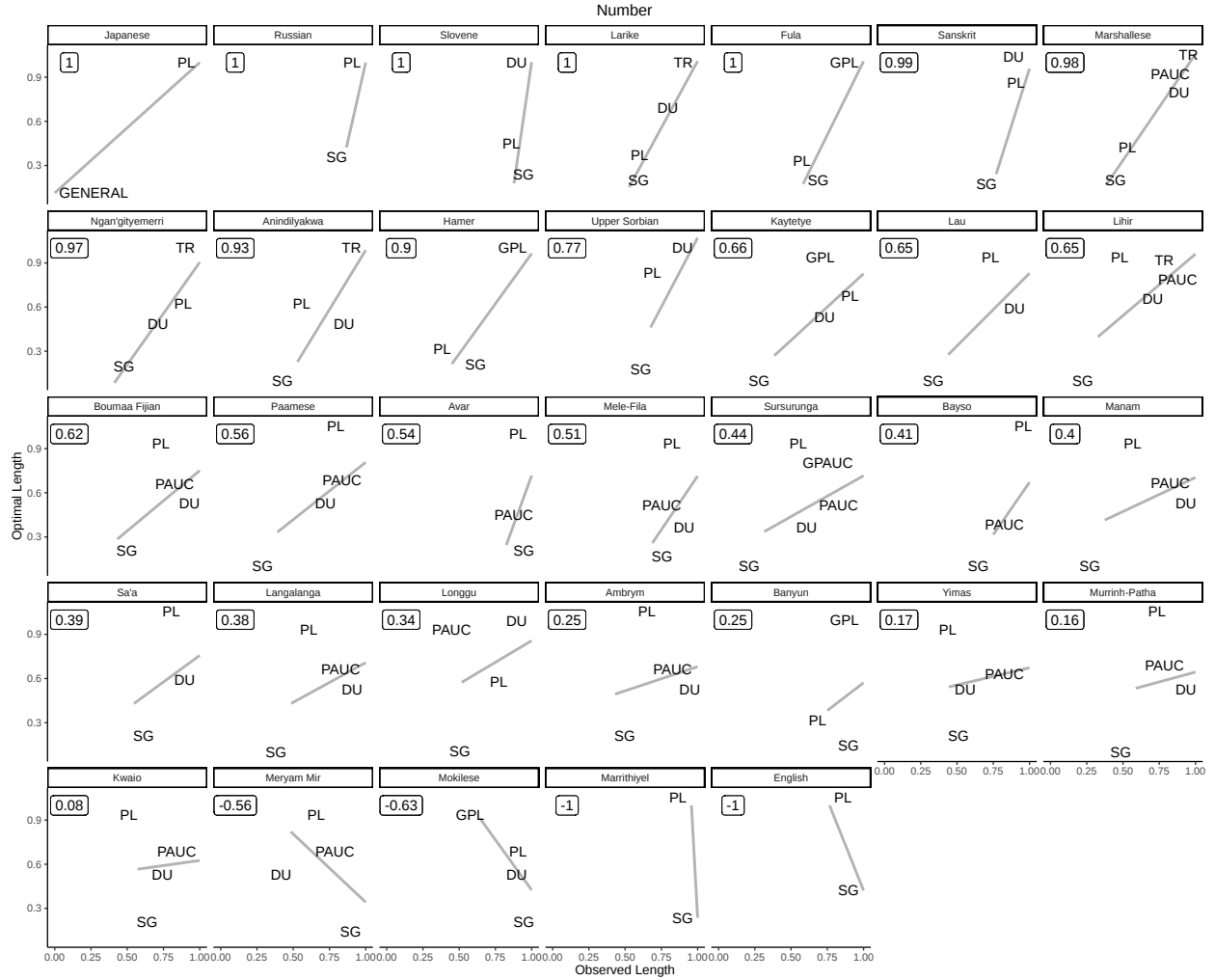
Figure S6: Optimal lengths against observed lengths for individual languages in our number data set. Correlations are shown at the top left of each panel.

S4 Individual Language Analyses

As we move from top left (minimal complexity) to bottom right (maximal complexity), along the IB Pareto frontier, the effective number of feature values in near-optimal systems gradually increases (Zaslavsky, 2020). Using the methods developed in Zaslavsky (2020), we can identify and analyze these structural changes as complexity is increased. This type of analysis was proposed and subsequently tested as a predictive model for the evolutionary trajectory of color naming systems (Zaslavsky et al., 2018; Zaslavsky, 2020). For example, the model recapitulates a set of well-known typological claims (Berlin & Kay, 1969), such as if a language has a category for green then it must have a category for red. This type of analysis has also been used to account for typological claims about an implicational evolutionary hierarchy of animal taxonomies (Zaslavsky, Regier, et al., 2019), suggesting that this theoretically-driven approach may apply more broadly to the evolution of the lexicon (Zaslavsky, 2020). In our context, similar typological statements have been made for the grammatical features in our analysis. For example, if a language makes a dual distinction in number it must make a singular and plural distinction. Traversing the IB Pareto frontier in our case will generate a theoretical trajectory for grammatical features. However, the optimal systems along this frontier are generally stochastic (Tishby et al., 1999), while systems for attested grammatical features are often deterministic. We will therefore also examine the deterministic systems for each number of distinctions that are the closest to

Figure S7: Optimal lengths against observed lengths for individual languages in our tense data set. Correlations are shown at the top left of each panel.

optimal systems along the frontier. We use gNID as the distance measure to identify the optimal system closest to each deterministic system.

In this section, we visualize several optimal systems, together with their deterministic approximations, at discrete points along the IB frontier, as the number of terms increases. We then qualitatively compare these theoretical systems with individual languages in our sample. In each case we omit discussion of the least complex system, which has a single feature value, and the most complex system, which has unique feature values for every state of the world. We connect these trajectories with typological claims about grammatical features, but make no strong claims about diachronic change given that our results are based on synchronic data. Although we organize our discussion around a sequence of systems that increase in complexity, we do not claim that evolutionary forces always act in this direction, and do not commit to any single mechanism that may serve to increase or decrease complexity. For example, there may be directional changes in the mode of expression (syntactic forms becoming morphological) or there may be changes in our assumptions about communicative need or speaker distributions over time and language contact. Closer modelling of the pathways of grammaticalization (e.g., as described by J. L. Bybee et al., 1994) is needed in future research.

For ease of exposition, we highlight clear categorical predictions of the model (e.g. the past tense should be subdivided before the future). These predictions, however, are subject to uncertainty about the assumptions made when applying the model. For example, the predicted asymmetry between past and future is induced

Figure S8: Optimal lengths against observed lengths for individual languages in our evidentiality data set.

by an asymmetry in usage probabilities for past (0.274) compared to future (0.251), and would reverse if these probabilities were exchanged. Several corpus analyses suggest that the past is mentioned more frequently than the future (Greenberg, 1966; Szagun, 1978), but other aspects of our approach have a less firm empirical grounding. In particular, as suggested earlier the hierarchy assumed for evidentiality and the prior over this hierarchy should both be viewed as provisional. This section lays out the fine-grained predictions that result from our best attempt to ground the theoretical framework in available empirical data, but some of these predictions may need to be revised as better characterizations of speaker distributions and usage frequencies become available.

S4.1 Number

Previous work on grammatical number has proposed a Number Hierarchy[3]:

$$\mathrm{SG} > \mathrm{PL} > \mathrm{DU} > \mathrm{TR}, \tag{9}$$

most likely hailing from Greenberg (1963, pg. 94)'s universal 34:

> "No language has a trial number unless it has a dual. No language has a dual unless it has a plural."

---

[3]We adopt our exposition and notation from Corbett (2000).

As the original universal does not mention the paucal, Croft (1990) and Foley (1986) have proposed to extend the hierarchy to:

$$\text{SG} > \text{PL} > \text{DU} > \text{TR}/\text{PAUC}. \tag{10}$$

Corbett (2000) argues that this would still be insufficient to explain the world's languages. Under (10), the presence of a paucal would imply the presence of a dual. This does not occur in Bayso, which distinguishes SG, PAUC, PL. Corbett (2000) also notes that number hierarchies make incorrect predictions when a feature value is optional. For example, in Serbian, the dual is optional so a speaker has the choice to use DU or PL morphology when talking about their *two* feet. For comparison, Sanskrit has the same "feature values" as Serbian; however, the speaker can only use the dual when talking about their *two* feet. Ostensibly under (10), an optional value should result in a choice between the value and its immediate predecessor in the hierarchy. While this works for Slovenian, it does not work for systems like Ngan'gityemerri, which distinguishes between SG, DU, TR and PL and has an optional trial. While (10) predicts a choice between TR and DU, the immediate predecessor, speakers of Ngan'gityemerri actually choose between TR and PL.

To account for these data, Corbett (2000) proposes a binary branching structure, where different languages "choose" how much they carve out of the plural. If a language carves out a determinate amount–i.e., exact value, it must follow the traditional number hierarchy (9). At any step, a language could choose to carve out an indeterminate amount (e.g., a paucal). Under this account, optional distinctions can be made by collapsing the lowest distinctions back into the plural. This collapse can occur multiple times in a language. For example, Larike has SG, DU, TR and PL values but both the dual and the trial can be used optionally with the plural. Using this typology, all of the data can be accommodated; however, the exact trajectory by which a language would increase its set of feature values is less precisely stated.

Instead of a binary branching structure, Harbour (2003, 2014) explains the diversity of attested number systems using a feature based approach and lattice theory. Under this account, numerical systems can be described in terms of three binary features reflecting computations on a lattice. While this account covers attested data, it differs from our account in two important respects. First, Harbour (2014)'s account does not handle optional distinctions as Corbett's account and our account do. Second, indefinite feature values (e.g., paucal) must have their boundaries set by additional socio-cultural conventions; whereas, our account derives probable boundaries from functional pressures. In contrast to the accounts of both Corbett and Harbour, our framework provides a probability distribution over possible systems rather than a generative language for attested systems.

By traversing the Pareto frontier in Figure 2, we can identify systems with increasing numbers of feature values that achieve optimal trade-offs between complexity and information loss. The left column of Figure S9 shows stochastic systems that lie directly on the Pareto frontier, and the right column shows the deterministic systems that are closest to these stochastic systems. Each line in each panel shows a different feature value. For comparison, Figure S10 shows encoders represented in our sample of attested systems, and a full list of systems is provided in Table S7. In line with all prior accounts, the theoretically optimal two term system captures the SG/PL distinction. Our data, however, includes one two-term language that does not make this distinction. Japanese distinguishes between GENERAL and PL, and although this system is not strictly optimal it lies extremely close to the Pareto frontier.

For systems with three terms, the number hierarchy allows a distinction between SG, DU and PL, and the binary branching typology allows an additional system that distinguishes SG, PAUC, and PL. Our analysis suggests that the optimal three term system splits the plural into an indefinite feature value and, if languages had a pressure towards deterministic systems the closest one distinguishes SG, DU and PL. In our 10 state space, the optimal stochastic system would be consistent with either a SG/PAUC/PL split as seen in Bayso or a SG/PL/GPL[4] split as seen in Fula. That being said, if we scaled up our analysis to a larger state space, it is unlikely that the indeterminate categories would map to a PL/GPL distinction, suggesting that this attested distinction does not arise due to the tradeoff between information loss and complexity. The other attested three term system is SG/DU/PL, aligning with the near optimal deterministic system. Languages with optional dual like Slovene are closer to optimal than languages with non-optional dual like Sanskrit because collapsing across the optional dual results in the optimal two term system. That being said, languages with dual are still close to the optimal trade-off.

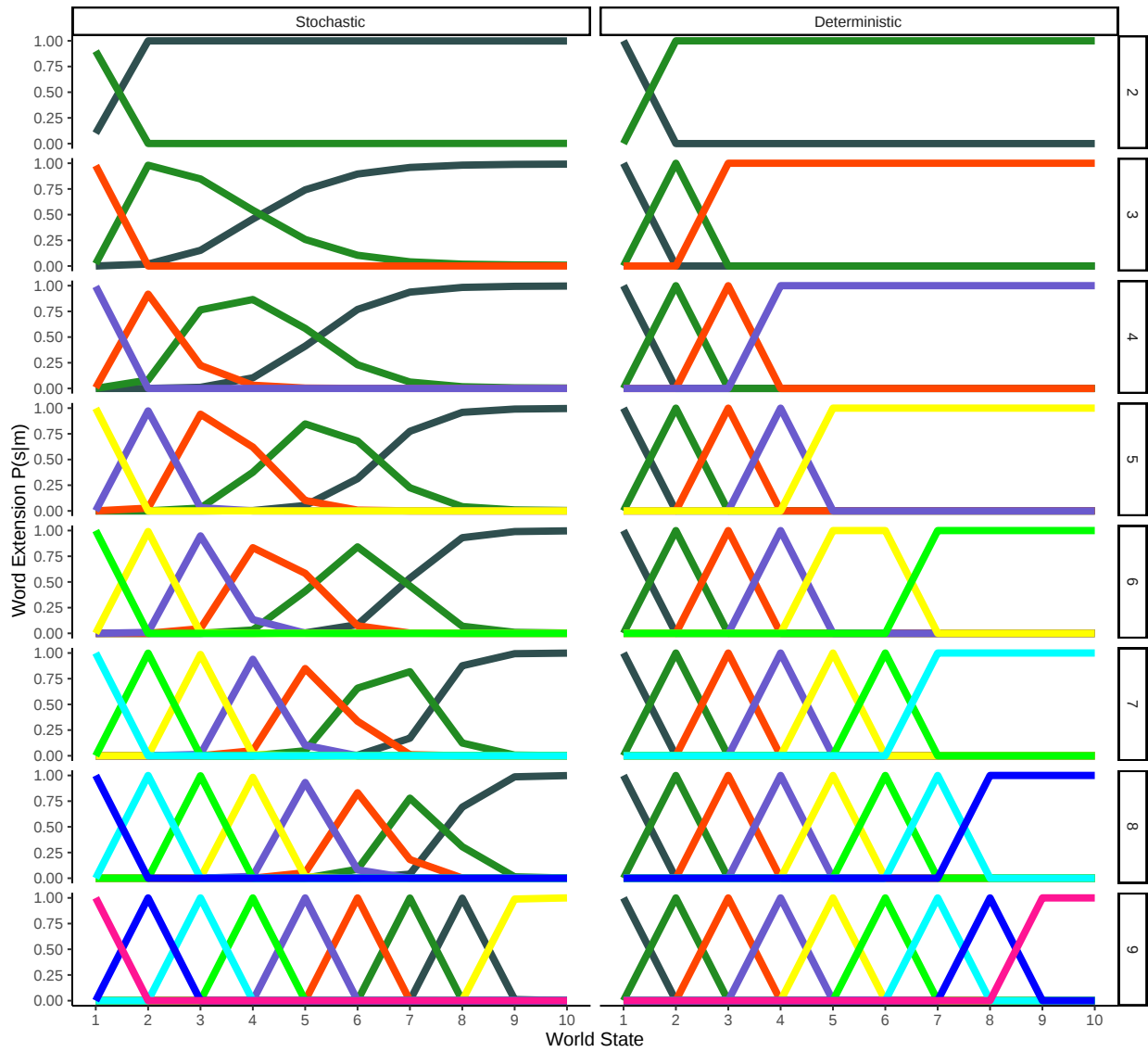[4]GPL stands for the greater plural, which corresponds to an exceptionally large number.

Figure S9: Number Encoder Systems. Optimal stochastic systems and most similar deterministic systems according to the model.
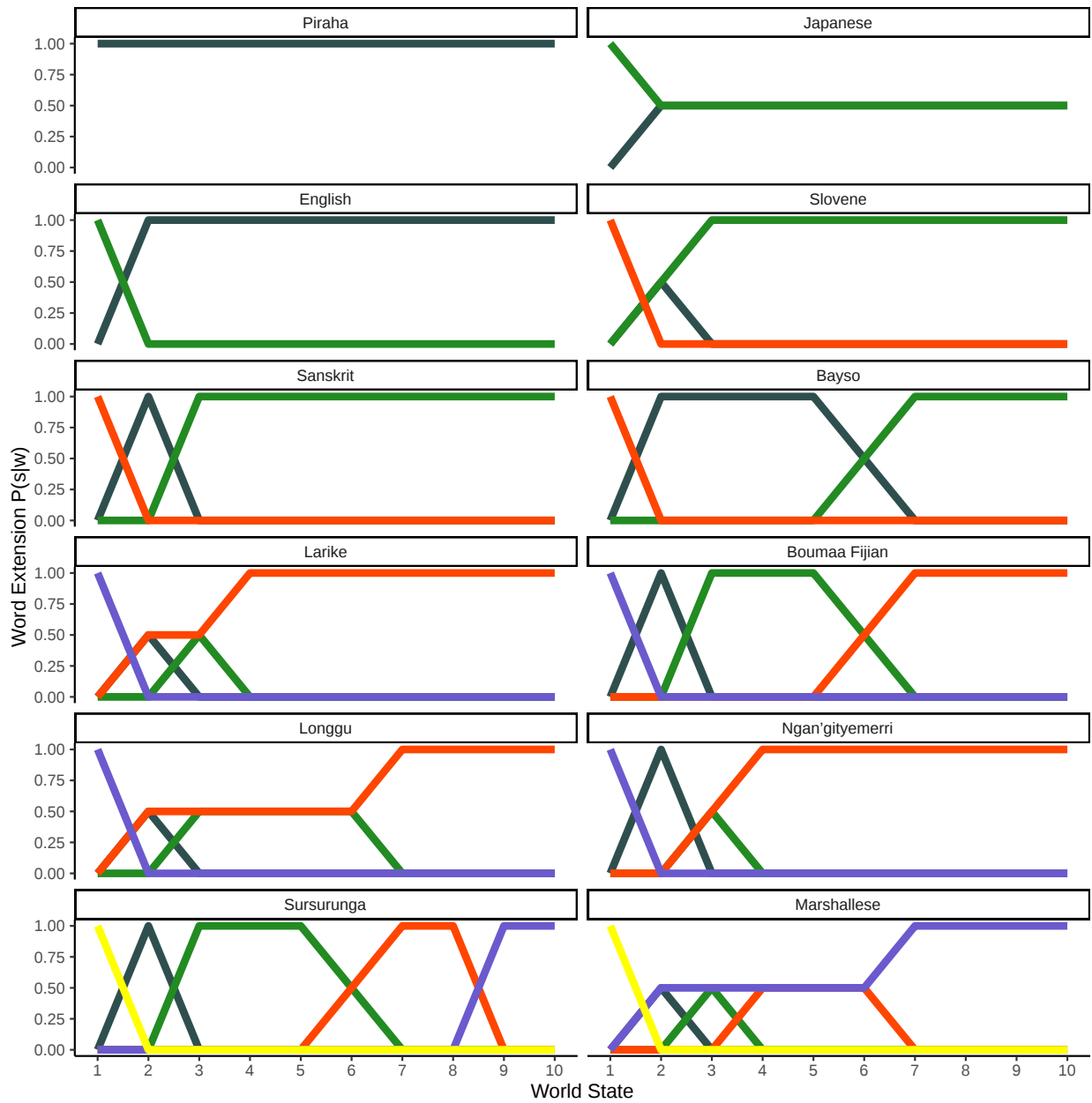
Figure S10: Number Encoder Systems. A sample of attested language systems in our coding scheme.

The optimal four term system according to our analysis is SG/DU/PAUC/PL, which is a very common system in Austronesian languages. There are a few other four term systems worth discussing. While making the same distinctions as the optimal system, both the dual and paucal in Longgu are optional. The resulting collapsed system would be the optimal two term system; however, the additional complexity of multiple optional terms pushes Longgu further away from the Pareto frontier. Similar to three term systems with two indeterminate categories, there are four term systems like Kaytetye which have a PL/GPL distinction instead of a PAUC/PL distinction. As for three term systems, it is again unclear whether four term systems with the PL/GPL distinction would continue to be near-optimal in an expanded state space. Other attested four term languages like Ngan'gityemerri split SG/DU/TR/PL, where the trial is always optional. In these cases, the collapsed system would not result in an optimal system. Larike makes the same split but with both an optional trial and an optional dual, resulting in an optimal collapsed system. Yet Larike is similar to Longgu in being pushed further from the frontier for having two optional distinctions. Recall, however, that all attested systems are still much closer to the Pareto frontier than the unattested systems in our analysis.

The optimal five term stochastic system splits another indeterminate feature value from plural SG/DU/PAUC/GPAUC/PL, which we see in Sursurunga. Even with three indeterminate categories, we still see at least one language that places a different range on those categories. Mele-Fila makes distinctions between SG/DU/PAUC/PL/GPL. Again it is unlikely this system would remain near-optimal if we scaled up our state space. While the optimal deterministic system in our analysis is SG/DU/TR/QUAD/PL, the second most similar system to optimal is SG/DU/TR/PAUC/PL which is exactly what we see in Lihir and Tangga. In our coding, we assumed that the trial in Lihir and Tangga is not optional. In truth, there seems to be a lack of data on the usage of the trial in these languages. Following the wider pattern of trials being optional, we know that Marshallese follows the optimal deterministic system but with optional dual, trial and paucal.

There are a few typological insights we can take away from our analysis. If languages were only trying to optimize the trade-off between information loss and complexity: 1) they should initially introduce an indeterminate distinction before each new determinate one; 2) determinate distinctions should be introduced according to the number hierarchy; 3) if a language favors minimizing complexity, it should introduce an optional determinate instead of an indeterminate; 4) if a language favors minimizing information loss, it should introduce a determinate. It appears that having an optional determinate distinction is less complex than having an indeterminate distinction; however, an additional indeterminate distinction reduces information loss more than an optional determinate.

As far as we know, there are no six term systems; however, under our account, we would predict a six term system to split SG/DU/TR/PAUC/GPAUC/PL or SG/DU/TR/QUAD/PAUC/PL. Given the added complexity of a quadral, we should not be surprised to see the quadral (and possibly also the trial and dual) to be optional. Given the pattern in our data, it would not be surprising if the GPAUC/PL split matches ranges that would be better described as PL/GPL; however, this is not predicted by our model, which suggests that some other pressure drives the evolution of such systems.

S4.2 Tense

Unlike number, we are not aware of any proposals for the "evolutionary" trajectory of tense systems. The majority of work on the grammaticalization of tense systems has focused on the evolution of the *forms* of tense markers, highlighting independent universal pathways for the grammatical forms (J. L. Bybee et al., 1994). That being said, Comrie (1985, pg. 85) points out the "general tendency of languages to have a better developed past than future system." Dahl (1985, pg. 125) further elaborates on this observation:

> "If there is one or more distinctions of remoteness in a [Tense-Mood-Aspect] system, and reference can be made to objective time measures, one of the distinctions will be between "more than one day away" and "not more than one day away".

In our framework, we would then predict that the past tense would subdivide immediate past before remote past tense. We return to this prediction after describing the optimal system trajectory under our analysis.

The upper panels of Figure S11 show stochastic systems and their closest deterministic counterparts that emerge as we traverse the Pareto frontier from simple to complex. The bottom panels show the unique encoders in our set of attested tense systems. We provide a full list of attested systems in our analysis in Table S8. The optimal two term system splits past (ABC) from non-past (RXYZ). While the majority of
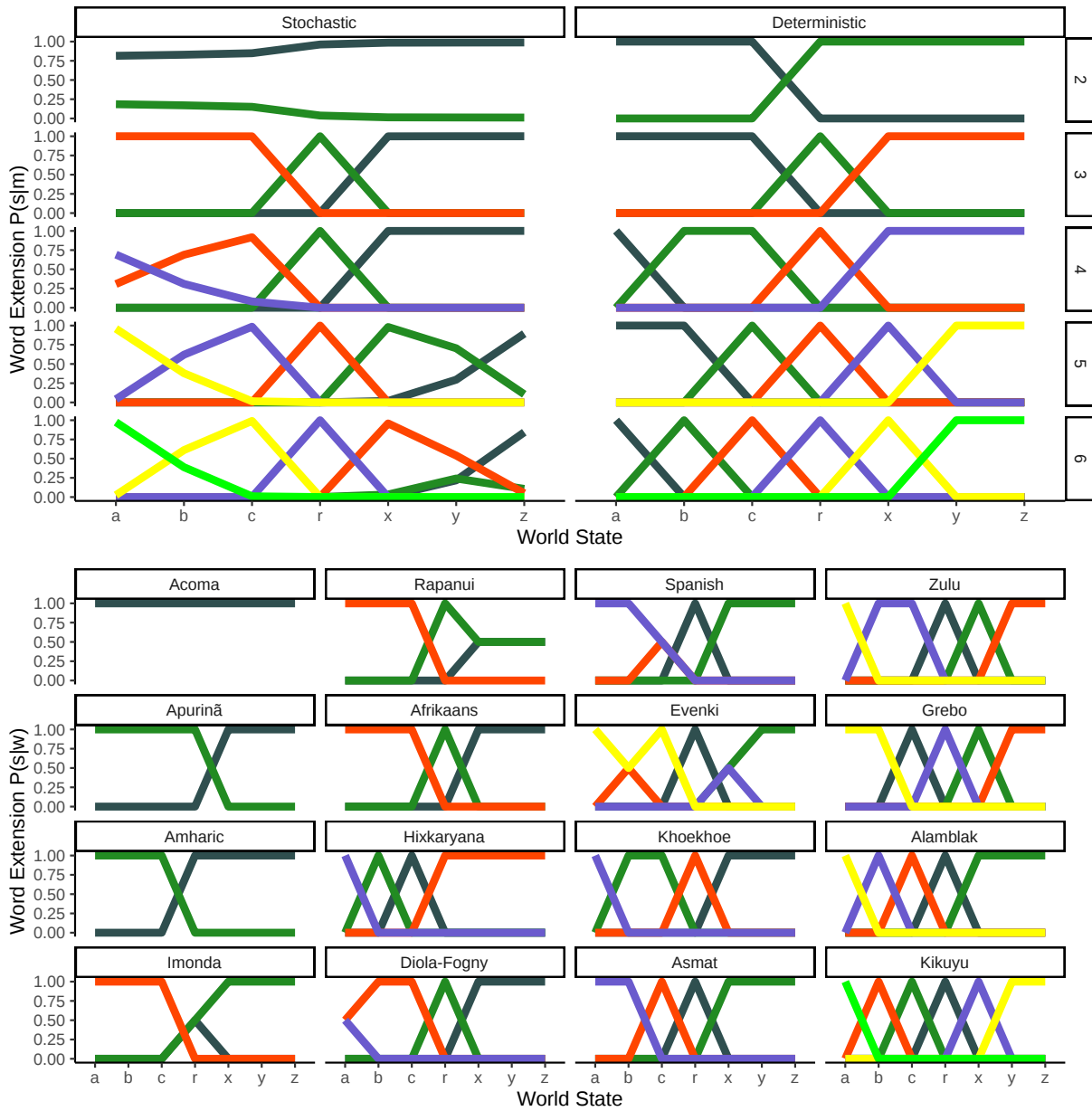
Figure S11: Tense Encoder Systems. Top: Optimal stochastic systems and most similar deterministic systems according to the model. Bottom: A sample of attested language systems in our coding scheme.

the two term systems in our analysis like Apurinã make this distinction, the remainder make a distinction between future (XYZ) and non-future (ABCR) like Amharic. In a larger sample than ours, Velupillai (2016b) finds more two term systems with a future vs non-future split than we do. She argues that grammatical aspect reduces the ambiguity between past and present such that an explicit tense for future is more helpful for communication. While our model does not currently handle multiple dimensions, future work can extend our approach to explore this correlation. In our framework, if complexity is weighted more heavily than information loss then present should be grouped with past, but if information loss carries more weight then present should be grouped with future. Both attested systems, however, lie close to the Pareto frontier.

Looking at three term systems, the optimal system divides the timeline according to the higher level structure in our meaning space: past (ABC), present (R) and future (XYZ). This system is also the most frequently attested three term system in our analysis. There are two exceptions. Imonda splits the timeline into past (ABC), present (R) and non-past (ABCR), and can be viewed as having an optional present. Rapanui splits the timeline into past (ABC), not-past (RXYZ) and future (XYZ), and can be viewed as a having an optional future. As we have seen with grammatical number systems, optional distinctions reduce the complexity of a given n-term system relative to a non-optional system; however, they often collapse to the optimal $n-1$ term system. Imonda and Rapanui both collapse to the near optimal two term systems and, thus, still lie close to the Pareto frontier.

Scaling up to four term systems, both the stochastic and deterministic models predict a subdivision of the past tense, as seen in Khoekhoe and Asmat. Spanish and Diola-Fogny make optional distinctions for immediate and remote past respectively, and, thus collapse to the optimal three-term system and lie close to the optimal frontier. Interestingly, one language in our sample uses a four-term system that lies relatively far from the Pareto frontier. Whereas the rest of the three term systems distinguish between future, present and two past tenses, Hixkaryana distinguishes between three past tenses and a non-past feature value. Retaining a non-past feature value instead of splitting into present and future results in a substantial information loss.

The optimal five term system splits into two past tenses, a present tense and two future tenses. Similar to the four term systems, some languages split the past tense into remote and general past like Zulu; whereas, other languages split the past tense into general and recent past like Grebo. Both types of attested optimal system always split the future into near future and general future. One five term system in our sample, Evenki, makes the optimal three term system split augmented with an optional general past (B) and an optional immediate future (X). Collapsing to the optimal three term system, Evenki still lies close to the Pareto frontier. Similar to Hixkaryana, we also find five term systems like Alamblak that carve up the past into three categories; however, these systems also split present tense and future tense, which mitigates the loss of communicative precision observed in Hixkaryana. As a result, they lie relatively close to the Pareto frontier.

The optimal deterministic six term system has three past tenses, a present tense and two future tenses; whereas, the optimal stochastic six term system has three futures, two pasts and a present tense. There is one language with six terms in our sample, Kikuyu. Kikuyu makes the same distinctions as the optimal deterministic six term system.

Looking at the patterns as a whole, if there is a present tense, there must be either a past or future tense. According to deterministic optimal systems, there should always be more past tenses than future tenses, suggesting that if there is an equal number of past and future tenses, the next step should be to subdivide the past. Interestingly, optimal deterministic systems always split off the remote before the immediate; whereas, optimal stochastic systems divide remoteness relatively evenly. As a result, systems that split off the immediate before the remote also lie very close to the Pareto frontier. Although our theory does not recapitulate Dahl's observation that immediate past takes priority over remote past, our timeline is relatively coarse. Future data sets may allow need probabilities and temporal uncertainties to be estimated more precisely, which may allow a better alignment between the theory and Dahl's observation.

S4.3 Evidentiality

To our knowledge, there are few claims about constraints on possible evidentiality systems, although Aikhenvald (2004) has suggested that complex systems correlate with small population sizes. While number and tense come with clear underlying dimensions, our meanings for evidentiality are more speculative. We know that humans do not automatically tag their memories with information source and that source monitoring ability
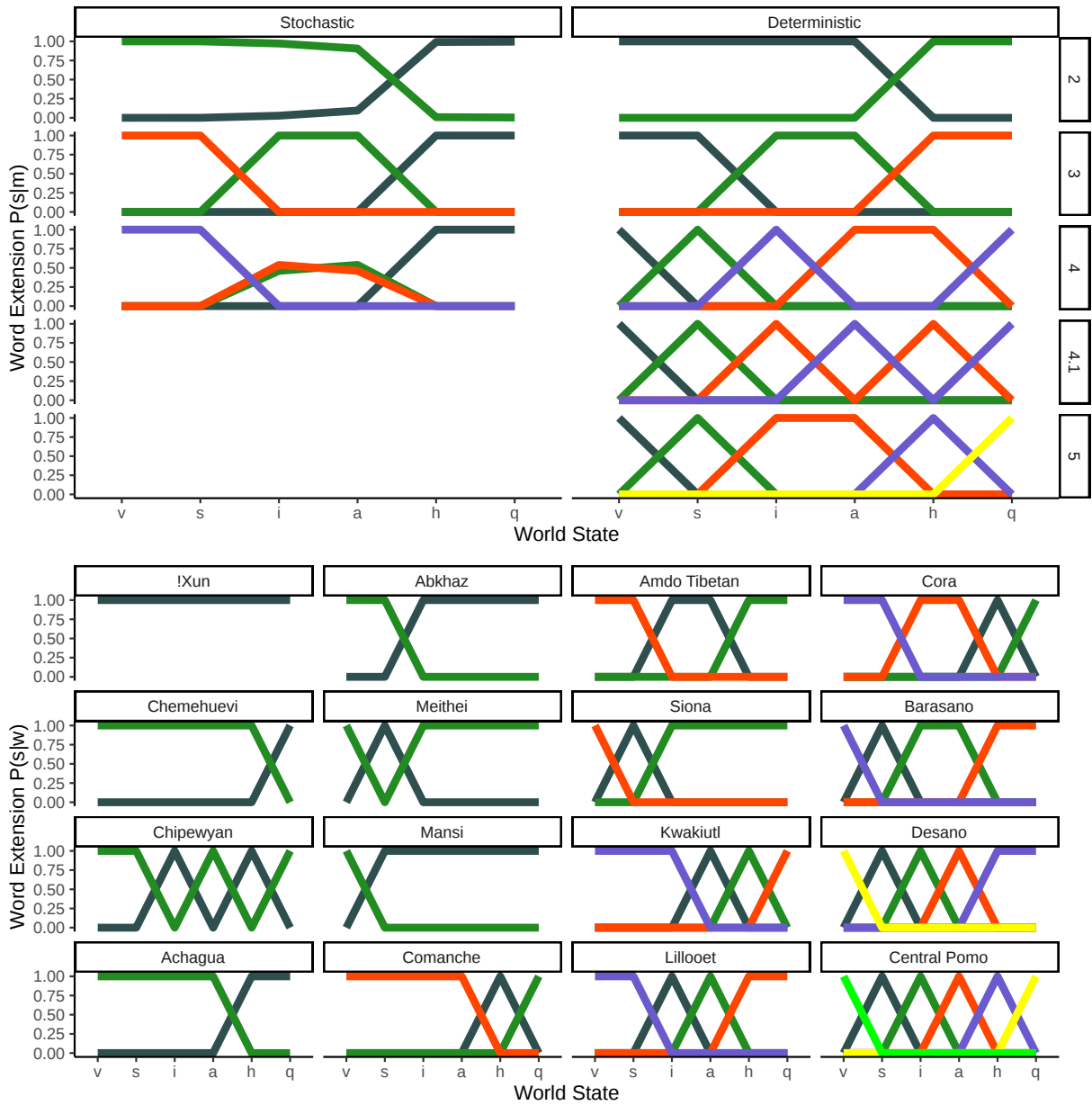
Figure S12: Evidentiality Encoder Systems. Top: Optimal stochastic systems and most similar deterministic systems according to the model. Bottom: A sample of attested language systems in our coding scheme.

is relatively poor among Western populations (Johnson et al., 1993; Saratsli et al., 2020). The speaker distributions in Figure 1 are therefore fairly broad, although constrained by the hierarchy over states that we have assumed. Given our uncertainty about the underlying semantic representation, our analysis of evidentiality should be viewed as more provisional than our analyses of number and tense.

The upper panels of Figure S12 show stochastic systems and their closest deterministic counterparts that emerge as we traverse the Pareto frontier from simple to complex. The bottom panels show the unique encoders in our set of attested evidentiality systems. A complete list of attested systems in our data set is provided in Table S9.

The optimal two-term system according to our model distinguishes reported events (HQ) from non-reported events. This distinction is generalized in Aikhenvald (2004) to firsthand vs. non-firsthand. In attested systems, inferred and assumed sources can be grouped with firsthand as in Achagua or with non-firsthand as in Abkhaz. In practice, both systems lie close to the Pareto frontier. There are several attested two term systems that do not follow this pattern. Looking first at systems with non-firsthand categories, some systems like Chemehuevi include a more specific quotative feature value (Q) instead of picking out reported events more generally. A few Athabaskan systems (e.g., Chipewyan) construct their non-firsthand feature value from hearsay and inference. Both systems break the hierarchy over the meaning space assumed by our model. For systems like Chemehuevi, a quotative feature value should not exist without a hearsay feature value. For systems like Chipewyan, the non-firsthand feature value should not exclude assumed and quoted sources. That being said, they still lie near the Pareto frontier because the wider non-firsthand feature value (IAHQ) has much lower need probability than the narrow firsthand feature value (VS).

Looking at attested systems with firsthand categories, some systems like Mansi narrow their firsthand feature value to only include visual information sources. Similarly, Meithei narrows its firsthand feature value to only include sensory information sources. In contrast to the non-firsthand narrowed systems, these systems lie quite far from the Pareto frontier because they break the meaning hierarchy where the need probability is high. A close look at these systems suggests that the meanings they convey are not adequately captured by the underlying evidentiality dimension assumed by our model. For Mansi, Northern Kanthy, Svan, and Turkish, evidentiality correlates with mirativity—i.e., a marker expressing surprise. For example, even after watching Captain America wield Thor's hammer, one can say "Cap can wield Mjölnir" with the non-firsthand marker because only Thor was thought to be capable of using the hammer, making Captain America's use surprising. Similarly, evidentiality in Nenets has a complex interaction with the mood system. In Meithei, evidentials interact with person, such that the non-firsthand marker can apply when the listener, rather than the speaker, lacks firsthand knowledge. For example, one can say "I have eaten" with a non-firsthand marker even though the speaker has firsthand knowledge of having eaten, because the listener does not know. We do not know enough about Megrelian and Xakas to discuss usage in these languages, but if we speculate based on geography or phylogeny, we would expect Megrelian and Xakas to behave like Svan and Turkish. While we appeal to correlations with other dimensions to explain these sub-optimal systems, correlations likely exist with other dimensions for all evidentiality systems, and for all other grammatical systems as well. Future analyses will therefore need to scale up to meaning spaces with multiple semantic dimensions in order to draw more definitive conclusions about constraints on linguistic diversity and the optimality of complete grammatical systems.

Moving on to three term systems, the optimal system in our model distinguishes between direct evidence (VS), indirect evidence (IA) and reported evidence (HQ), again borrowing Aikhenvald (2004)'s terminology. The majority of three term systems, including Amdo Tibetan and Quechua, follow the optimal system. A few languages do not make the split along the top level of our meaning hierarchy and instead further split a reported evidential into two categories. For example, Comanche has a general quotative marker and adopts a marker reflecting information that the speaker does not know in narrative discourse. The last three term system in our analysis, Siona, splits visual, sensory and non-firsthand sources. Similar to Mansi, our analysis suggests that splitting visual and sensory categories at this stage would be sub-optimal. However, closer analysis of Siona shows that evidentiality correlates with the degree of control the speaker had over the situation. Full control receives visual marking, partial control receives sensory marking and no control receives non-firsthand marking. While these systems do not make all top level distinctions in the meaning hierarchy, the distinctions they implement still respect the hierarchy and so they do lie relatively close to the Pareto frontier.

As meanings are not strongly biased within each hierarchical cluster, the optimal stochastic systems

do not make strong predictions for four and five term systems. The optimal four term system weakly subdivides the indirect evidence feature value into assumed and inferred sources, as seen deterministically in Lillooet. Other attested systems, such as Cora, subdivide the reported feature value into hearsay and quotative sources instead. Others, such as Barasano, subdivide the direct evidence feature value into visual and sensory sources. These systems all respect the meaning hierarchy and, thus, lie near the Pareto frontier. One language, Kwakiutl, has firsthand (VSI), assumed (A), hearsay (H) and quotative (Q) markers, which violates the assumed meaning hierarchy. This violation, however, occurs in a part of the hierarchy with low need probability, meaning that the system does not lie too far from the Pareto frontier.

The bias in meanings and relative frequency between visual/sensory and hearsay/quotative are mathematically equivalent. Both distinctions therefore emerge at the same rate, and there is no optimal five term system. The five term systems in our sample (e.g., Desano) fully subdivide direct and indirect evidence while leaving the reported evidence intact. Again this system obeys the meaning hierarchy and, therefore, lies along the Pareto frontier.

The overall pattern with our evidentiality analysis is that distance to the Pareto frontier corresponds to how many distinctions in the meaning hierarchy are preserved. This finding corresponds nicely with Gentner and Bowerman (2009)'s proposal that the naturalness of a concept corresponds to its typological frequency. Following the meaning hierarchy, optimal systems divide the top level clusters before making finer distinctions. When choosing to divide hierarchical clusters at the same level, the model predicts dividing the feature value with the lowest need probability. This prediction is consistent with Dahl (1985)'s intuition that all else being equal, less frequent categories should be preferred in use over a broad frequent feature value; otherwise, the less frequent feature value would never be used at all.

Language (Source)	System	Complexity	Info. Loss	Frontier Dist.	gNID	Tradeoff
Piraha (Corbett, 2000—Corbett, 2000: 50-51)	GENERAL	0.00	1.44	0.00	0.00	1.02
Kawi (Corbett, 2000—Corbett, 2000: 51)	GENERAL	0.00	1.44	0.00	0.00	1.02
Japanese (Corbett, 2000—Corbett, 2000: 14)	GENERAL/PL	0.32	1.14	0.01	-0.24	1.03
English (Corbett, 2000—Corbett, 2000: 9)	SG/PL	0.94	0.55	0.01	0.06	1.16
Russian (Corbett, 2000—Corbett, 2000: 4)	SG/PL	0.94	0.55	0.01	0.06	1.16
Slovene (Corbett, 2000—Corbett, 2000: 43-45)	SG/DU$_O$/PL	1.05	0.45	0.01	0.03	1.17
Bayso (Corbett, 2000—Corbett, 2000: 22)	SG/PAUC/PL	1.12	0.40	0.01	0.06	1.24
Avar (Corbett, 2000—Corbett, 2000: 22)	SG/PAUC/PL	1.12	0.40	0.01	0.06	1.24
Banyun (Corbett, 2000—Corbett, 2000: 31)	SG/PL/GPL	1.12	0.40	0.01	0.06	1.24
Fula (Corbett, 2000—Corbett, 2000: 31)	SG/PL/GPL	1.12	0.40	0.01	0.06	1.24
Hamer (Corbett, 2000—Corbett, 2000: 32-33)	SG/PL/GPL	1.12	0.40	0.01	0.06	1.24
Longgu (Corbett, 2000—Corbett, 2000: 46-48)	SG/DU$_O$/PAUC$_O$/PL	1.13	0.40	0.02	0.05	1.20
Larike (Corbett, 2000—Corbett, 2000: 21)	SG/DU$_O$/TR$_O$/PL	1.13	0.40	0.03	0.04	1.19
Marshallese (Corbett, 2000—Corbett, 2000: 29-30)	SG/DU$_O$/TR$_O$/PAUC$_O$/PL	1.21	0.35	0.04	0.05	1.23
Upper Sorbian (Corbett, 2000—Corbett, 2000: 20)	SG/DU/PL	1.29	0.26	0.02	0.05	1.47
Sanskrit (Corbett, 2000—Corbett, 2000: 45)	SG/DU/PL	1.29	0.26	0.02	0.05	1.47
Ngan'gityemerri (Corbett, 2000—Corbett, 2000: 21-22;43-44)	SG/DU/TR$_O$/PL	1.35	0.22	0.02	0.03	1.50
Marrithiyel (Corbett, 2000—Corbett, 2000: 22;43)	SG/DU/TR$_O$/PL	1.35	0.22	0.02	0.03	1.50
Anindilyakwa (Corbett, 2000—Corbett, 2000: 22;43)	SG/DU/TR$_O$/PL	1.35	0.22	0.02	0.03	1.50
Yimas (Corbett, 2000—Corbett, 2000: 23)	SG/DU/PAUC/PL	1.43	0.16	0.01	0.02	1.71
Meryam Mir (Corbett, 2000—Corbett, 2000: 23)	SG/DU/PAUC/PL	1.43	0.16	0.01	0.02	1.71
Boumaa Fijian (Corbett, 2000—Corbett, 2000: 23-24)	SG/DU/PAUC/PL	1.43	0.16	0.01	0.02	1.71
Paamese (Corbett, 2000—Corbett, 2000: 24)	SG/DU/PAUC/PL	1.43	0.16	0.01	0.02	1.71
Manam (Corbett, 2000—Corbett, 2000: 24)	SG/DU/PAUC/PL	1.43	0.16	0.01	0.02	1.71
Ambrym (Corbett, 2000—Corbett, 2000: 24)	SG/DU/PAUC/PL	1.43	0.16	0.01	0.02	1.71
Kwaio (Corbett, 2000—Corbett, 2000: 25)	SG/DU/PAUC/PL	1.43	0.16	0.01	0.02	1.71
Sa'a (Corbett, 2000—Corbett, 2000: 25)	SG/DU/PAUC/PL	1.43	0.16	0.01	0.02	1.71
Langalanga (Corbett, 2000—Corbett, 2000: 25)	SG/DU/PAUC/PL	1.43	0.16	0.01	0.02	1.71
Lau (Corbett, 2000—Corbett, 2000: 25)	SG/DU/PAUC/PL	1.43	0.16	0.01	0.02	1.71
Ungarinjin (Corbett, 2000—Corbett, 2000: 25)	SG/DU/PAUC/PL	1.43	0.16	0.01	0.02	1.71
Murrinh-Patha (Corbett, 2000—Corbett, 2000: 25)	SG/DU/PAUC/PL	1.43	0.16	0.01	0.02	1.71
Kaytetye (Corbett, 2000—Corbett, 2000: 33)	SG/DU/PL/GPL	1.43	0.16	0.01	0.02	1.71
Mokilese (Corbett, 2000—Corbett, 2000: 34)	SG/DU/PL/GPL	1.43	0.16	0.01	0.02	1.71
Sursurunga (Corbett, 2000—Corbett, 2000: 26-27)	SG/DU/PAUC/GPAUC/PL	1.47	0.14	0.02	0.04	1.84
Mele-Fila (Corbett, 2000—Corbett, 2000: 35)	SG/DU/PAUC/PL/GPL	1.47	0.14	0.02	0.04	1.84
Lihir (Corbett, 2000—Corbett, 2000: 25-26)	SG/DU/TR/PAUC/PL	1.57	0.07	0.01	0.01	2.37
Tangga (Corbett, 2000—Corbett, 2000: 29)	SG/DU/TR/PAUC/PL	1.57	0.07	0.01	0.01	2.37

Table S7: Attested grammatical number systems and their parameters according to the model. The trade-off parameter, β, provided corresponds to the beta value that maximizes similarity. SG-singular; PL-plural; DU-dual; TR-trial; PAUC-paucal; GPAUC-greater paucal; GPL-greater plural. $_O$ denotes optional.

Language (Source)	System	Complexity	Info. Loss	Frontier Dist.	gNID	Tradeoff
Acoma (Miller, 1965—Miller, 1965)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Ainu (Refsing, 1986—Refsing, 1986)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Amele (Roberts, 1987—Roberts, 1987: 227-228)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Araona (Pitman, 1980—Pitman, 1980)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Arapesh (Mountain; Conrad & Wogiga, 1991—Conrad & Wogiga, 1991)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Bagirmi (Stevenson, 1969—Stevenson, 1969)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Berber (Middle Atlas; Penchoen, 1973—Penchoen, 1973)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Cebuano (Dahl, 1985—Dahl, 1985: 160)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Cree (Plains; Wolfart, 1996—Wolfart, 1996)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Dani (Lower Grand Valley; Bromley, 1981—Bromley, 1981: 61-65)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Garífuna (J. L. Bybee et al., 1994—Bybee et al., 1994)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Hausa (Newman, 2000—Newman, 2000)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Hmong Njua (Harriehausen, 1990—Harriehausen, 1990)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Jakaltek (Grinevald Craig, 1977—Grinevald Craig, 1977)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Ju|'hoan (Dickens, 1991—Dickens, 1991)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Karok (J. L. Bybee et al., 1994—Bybee et al., 1994)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Ket (Krejnovich, 1968—Krejnovich, 1968)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Khmer (Jacob, 1968—Jacob, 1968)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Kiowa (Watkins, 1984—Watkins, 1984)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Koyraboro Senni (Heath, 1999—Heath, 1999)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Kutenai (Boas, 1927—Canestrelli & Boas, 1927)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Lahu (J. L. Bybee et al., 1994—Bybee et al., 1994)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Lakhota (Buechel, 1939—Buechel, 1939)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Lango (Noonan, 1992—Noonan, 1992)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Lavukaleve (Terrill, 1999—Terrill, 1999)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Luganda (M. Dryer & Haspelmath, 2020—WALS)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Luvale (Horton, 1949—Horton, 1949)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Malagasy (Dez, 1980—Dez, 1980)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Mapudungun (Smeets, 1989—Smeets, 1989)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Maricopa (Gordon, 1986—Gordon, 1986)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Maybrat (Dol, 1999—Dol, 1999)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Meithei (Chelliah, 1997—Chelliah, 1997)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Motu (J. L. Bybee et al., 1994—Bybee et al., 1994)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Ngambay (J. L. Bybee et al., 1994—Bybee et al., 1994)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Nicobarese (Car; J. L. Bybee et al., 1994—Bybee et al. 1994)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Paiwan (Chang, 2006—Chang, 2006)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Pirahã (Everett, 1986—Everett, 1986)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Rendille (M. Dryer & Haspelmath, 2020—WALS)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Sango (Thornell, 1997—Thornell, 1997)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Sanuma (Borgman, 1999—Borgman, 1990: 167)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Semelai (Kruspe, 2004—Kruspe, 2004)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Suena (Wilson, 1974—Wilson, 1974: 39-40)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Tagalog (Schachter & Otanes, 1972—Schachter & Otanes, 1972)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Tukang Besi (Donohue, 1999—Donohue, 1999)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Vietnamese (Nguyen, 1997—Nguyen, 1997)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Yagua (Payne, 1990—Payne & Payne, 1990: 386-87)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Yucatec (Bohnemeyer, 2002—Bohnemeyer, 2002)	(ABCRXYZ)	0.00	1.01	0.00	0.00	1.29
Apurinã (Facundes, 2000—Facundes, 2000: 517)	(ABCR)(XYZ)	0.81	0.50	0.04	0.27	1.80
Canela-Krahô (Popjes & Popjes, 1986—Popjes & Popjes, 1986)	(ABCR)(XYZ)	0.81	0.50	0.04	0.27	1.80
Chamorro (Topping & Dungca, 1973—Topping, 1973)	(ABCR)(XYZ)	0.81	0.50	0.04	0.27	1.80
Greenlandic (West; Dahl, 1985—Dahl, 1985: 165)	(ABCR)(XYZ)	0.81	0.50	0.04	0.27	1.80
Indonesian (Dahl, 1985—Dahl, 1985: 160)	(ABCR)(XYZ)	0.81	0.50	0.04	0.27	1.80
Isekiri (Dahl, 1985—Dahl, 1985: 178)	(ABCR)(XYZ)	0.81	0.50	0.04	0.27	1.80
Javanese (Dahl, 1985—Dahl, 1985: 161)	(ABCR)(XYZ)	0.81	0.50	0.04	0.27	1.80
Kayardild (N. D. Evans, 1995—Evans, 1995)	(ABCR)(XYZ)	0.81	0.50	0.04	0.27	1.80
Khmu' (Dahl, 1985—Dahl, 1985: 174)	(ABCR)(XYZ)	0.81	0.50	0.04	0.27	1.80
Lai (J. L. Bybee et al., 1994—Bybee et al., 1994)	(ABCR)(XYZ)	0.81	0.50	0.04	0.27	1.80
Mandarin (Dahl, 1985—Dahl, 1985: 180)	(ABCR)(XYZ)	0.81	0.50	0.04	0.27	1.80
Nivkh (Panfilov, 1962—Panfilov, 1962)	(ABCR)(XYZ)	0.81	0.50	0.04	0.27	1.80
Oneida (Dahl, 1985—Dahl, 1985: 173)	(ABCR)(XYZ)	0.81	0.50	0.04	0.27	1.80
Seneca (Dahl, 1985—Dahl, 1985: 173)	(ABCR)(XYZ)	0.81	0.50	0.04	0.27	1.80
Thai (Dahl, 1985—Dahl, 1985: 174)	(ABCR)(XYZ)	0.81	0.50	0.04	0.27	1.80
Wari' (Everett & Kern, 1997—Everett & Kern, 1997)	(ABCR)(XYZ)	0.81	0.50	0.04	0.27	1.80
Yoruba (Dahl, 1985—Dahl, 1985: 179)	(ABCR)(XYZ)	0.81	0.50	0.04	0.27	1.80
Yukaghir (Kolyma; Maslova, 2003—Maslova, 1999)	(ABCR)(XYZ)	0.81	0.50	0.04	0.27	1.80
Amharic (Dahl, 1985—Dahl, 1985: 156)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Arabic (Egyptian; Eisele, 1999—Eisele, 1999)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Arabic (Tunisian; Dahl, 1985—Dahl, 1985: 155)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Avokaya (M. Dryer & Haspelmath, 2020—WALS)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Awa Pit (Curnow, 1997—Curnow, 1997: 176)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Brahui (Andronov, 1980—Andronov, 1980)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Burushaski (Berger, 1998—Berger, 1998: 142, 160-161)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Daga (Murane, 1974—Murane, 1974: 53)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Fijian (Dixon, 1988—Dixon, 1988: 73)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Finnish (Dahl, 1985—Dahl, 1985: 181)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Inuktitut (Salluit; Hayashi & Spreng, 2005—Hayashi & Spreng, 2005)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Japanese (Dahl, 1985—Dahl, 1985: 158)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Khalkha (Svantesson, 1991—Svantesson, 1991)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Korean (Lee, 1991—Lee, 1991: 225)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Lezgian (Haspelmath, 1993—Haspelmath, 1993: 146)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Mangarrayi (Merlan, 1982—Merlan, 1982: 61)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Maori (Dahl, 1985—Dahl, 1985: 163)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Martuthunira (Dench, 1995—Dench, 1995: 144)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Nenets (Salminen, 1997—Salminen, 1997)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Oromo (Harar; Owens, 1985—Owens, 1985: 73)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Rama (Grinevald, 1990—Grinevald, 1990)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Sundanese (Dahl, 1985—Dahl, 1985: 139)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Tiwi (Osborne, 1974—Osborne, 1974: 40)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Warao (Romero-Figueroa, 1997—Romero-Figueroa, 1997: 95)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Yaqui (Dedrick & Casad, 1999—Dedrick & Casad, 1999: 318)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Zoque (Copainalá; Harrison, Harrison, & García, 1981—Harrison et al., 1981: 92)	(ABC)(RXYZ)	0.85	0.47	0.04	0.25	1.82
Imonda (Seiler, 1985—Seiler, 1985)	(ABC)(R)(RXYZ)	1.03	0.38	0.06	0.10	1.85
Rapanui (Du Feu, 1996—Du Feu, 1996: 156)	(ABC)(RXYZ)(XYZ)	1.08	0.32	0.03	0.09	1.88
Hixkaryana (Derbyshire, 1979—Derbyshire, 1979: 138-139)	(A)(B)(C)(RXYZ)	1.26	0.42	0.21	0.38	2.36
Afrikaans (Dahl, 1985—Dahl, 1985: 166)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Akan (Dahl, 1985—Dahl, 1985: 179)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Alawa (Dahl, 1985—Dahl, 1985: 159)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Armenian (Eastern; Dum-Tragut, 2009—Dum-Tragut, 2009)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Azerbaijani (Dahl, 1985—Dahl, 1985: 159-160)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Bandjalang (Dahl, 1985—Dahl, 1985: 159)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Beja (Dahl, 1985—Dahl, 1985: 154-155)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Bengali (Dahl, 1985—Dahl, 1985: 167)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Bulgarian (Dahl, 1985—Dahl, 1985: 172)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Chukchi (M. J. Dunn, 1999—Dunn, 1999)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Didinga (Lohitare, Lohammarimoi, Peter, & Joseph, 2012—Lohitare et al., 2012)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
English (Dahl, 1985—Dahl, 1985: 166)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
French (Dahl, 1985—Dahl, 1985: 170)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Georgian (Dahl, 1985—Dahl, 1985: 164)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
German (Dahl, 1985—Dahl, 1985: 166)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Gooniyandi (McGregor, 1990—McGregor, 1990: 20)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Greek (Modern; Dahl, 1985—Dahl, 1985: 169)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Guaraní (Dahl, 1985—Dahl, 1985: 159)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Hebrew (Modern; Dahl, 1985—Dahl, 1985: 114)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Hindi (Dahl, 1985—Dahl, 1985: 168-169)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Hungarian (Dahl, 1985—Dahl, 1985: 181)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Icelandic (Svenonius, 2002—Svenonius, 2002)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Kannada (Sridhar, 1990—Sridhar, 1990: 219)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Kriol (Fitzroy Crossing; Dahl, 1985—Dahl, 1985: 166)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Kurmanji (Dahl, 1985—Dahl, 1985: 168)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Latvian (Hewson & Bubenik, 1997—Hewson & Bubenik, 1987)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Makah (Davidson, 2002—Davidson, 2002)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Maltese (Dahl, 1985—Dahl, 1985: 156)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Mixtec (Chalcatongo; Macaulay, 1996—Macaulay, 1996)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95

Montagnais (Drapeau, 2014—Drapeau, 2014)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Ngiyambaa (Donaldson, 1980—Donaldson, 1980: 160)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Punjabi (Dahl, 1985—Dahl, 1985: 168)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Persian (Dahl, 1985—Dahl, 1985: 168)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Portuguese (Dahl, 1985—Dahl, 1985: 171-172)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Quechua (Cochabamba; Dahl, 1985—Dahl 1985: 158)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Quechua (Imbabura; Dahl, 1985—Dahl 1985: 144)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Romanian (Dahl, 1985—Dahl, 1985: 171-172)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Russian (Dahl, 1985—Dahl, 1985: 172)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Somali (Saeed, 1999—Saeed, 1999)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Swahili (Lindfors, 2003—Lindfors, 2003)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Swedish (Dahl, 1985—Dahl, 1985: 166)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Tamil (Dahl, 1985—Dahl, 1985: 165-166)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Tenyer (Dahl, 1985—Dahl, 1985: 177)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Tigrinya (Dahl, 1985—Dahl, 1985: 156)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Turkish (Dahl, 1985—Dahl, 1985: 157)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Wichita (Rood, 1976—Rood, 1976)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Wolaytta (Lamberti & Sottile, 1997—Lamberti & Sottile, 1997)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Wolof (Dahl, 1985—Dahl, 1985: 180)	(ABC)(R)(XYZ)	1.52	0.08	0.00	0.00	4.95
Diola-Fogny (Sapir, 1965—Sapir, 1965: 30-35)	(A)(ABC)(R)(XYZ)	1.61	0.07	0.00	0.00	5.53
Kewa (Franklin, 1971—Franklin, 1971)	(A)(ABC)(R)(XYZ)	1.61	0.07	0.00	0.00	5.53
Spanish (Dahl, 1985—Dahl, 1985: 171)	(ABC)(C)(R)(XYZ)	1.61	0.07	0.00	0.00	5.41
Evenki (Nedjalkov, 1997—Nedjalkov, 1997: 235)	(ABC)(B)(R)(X)(XYZ)	1.68	0.07	0.02	0.02	5.84
Khoekhoe (Hagman, 1977—Hagman, 1977: 64)	(A)(BC)(R)(XYZ)	1.76	0.05	0.01	0.05	10.40
Asmat (Voorhoeve, 1965—Voorhoeve, 1965: 98)	(AB)(C)(R)(XYZ)	1.80	0.05	0.01	0.06	11.03
Sesotho (Dahl, 1985—Dahl, 1985: 173)	(AB)(C)(R)(XYZ)	1.80	0.05	0.01	0.06	11.03
Alamblak (Bruce, 1984—Bruce, 1984: 133)	(A)(B)(C)(R)(XYZ)	1.93	0.04	0.01	0.06	13.19
Barasano (Jones & Jones, 1991—Jones & Jones, 1991: 84-86)	(A)(B)(C)(R)(XYZ)	1.93	0.04	0.01	0.06	13.19
Koasati (Kimball, 1985—Kimball, 1991: 203)	(A)(B)(C)(R)(XYZ)	1.93	0.04	0.01	0.06	13.19
Wichí (Wdzenczny, n.d.—Wdzenczny)	(A)(B)(C)(R)(XYZ)	1.93	0.04	0.01	0.06	13.19
Zulu (Dahl, 1985—Dahl, 1985: 175)	(A)(BC)(R)(X)(YZ)	2.01	0.02	0.00	0.05	14.61
Grebo (Innes, 1966—Innes, 1966: 54-87)	(AB)(C)(R)(X)(YZ)	2.05	0.02	0.00	0.05	15.45
Supyire (Carlson, 1994—Carlson, 1994: 329-334)	(AB)(C)(R)(X)(YZ)	2.05	0.02	0.00	0.05	15.45
Kikuyu (Dahl, 1985—Dahl, 1985: 175)	(A)(B)(C)(R)(X)(YZ)	2.18	0.01	0.00	0.03	19.66

Table S8: Attested tense systems and their parameters according to the model. A-remote past; B-past; C-near past; R-present; X-immediate future; Y-future; Z-remote future.

Language	System	Complexity	Info. Loss	Frontier Dist.	gNID	Beta
!Xun (Aikhenvald, 2004—Aikhenvald, 2004: 149)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Adioukrou (Aikhenvald, 2004—Aikhenvald 2004: 135)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Afghan Persian (Aikhenvald, 2004—Aikhenvald 2004: 109)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Aguaruna (Aikhenvald, 2004—Aikhenvald 2004: 137)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Andean Spanish (Aikhenvald, 2004—Aikhenvald 2004: 21)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Arabic (Aikhenvald, 2004—Aikhenvald 2004: 10)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Avar (Aikhenvald, 2004—Aikhenvald 2004: 38)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Chukchi (Aikhenvald, 2004—Aikhenvald 2004)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Colombian Spanish (Aikhenvald, 2004—Aikhenvald 2004: 141)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Dani (Aikhenvald, 2004—Aikhenvald 2004)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Donno Saw (Aikhenvald, 2004—Aikhenvald 2004: 133)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Dutch (Aikhenvald, 2004—Aikhenvald 2004: 17)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Dyirbal (Aikhenvald, 2004—Aikhenvald 2004: 132)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
English (Aikhenvald, 2004—Aikhenvald 2004: 10)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
French (Aikhenvald, 2004—Aikhenvald 2004: 17)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Georgian (Aikhenvald, 2004—Aikhenvald 2004: 38)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
German (Aikhenvald, 2004—Aikhenvald 2004)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Greek (Aikhenvald, 2004—Aikhenvald 2004: 150)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Hebrew (Aikhenvald, 2004—Aikhenvald 2004: 10)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Irantxe (Aikhenvald, 2004—Aikhenvald 2004: 75)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Ishkashim (Aikhenvald, 2004—Aikhenvald 2004: 38)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Istanbul Judezmo (Aikhenvald, 2004—Aikhenvald 2004: 114)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Japanese (Aikhenvald, 2004—Aikhenvald 2004: 81)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Kambera (Aikhenvald, 2004—Aikhenvald 2004: 140-141)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Ket (Aikhenvald, 2004—Aikhenvald 2004: 290)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Korean (Aikhenvald, 2004—Aikhenvald 2004)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Makah (Aikhenvald, 2004—Aikhenvald 2004)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Nivkh (Aikhenvald, 2004—Aikhenvald 2004: 290)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Panare (Aikhenvald, 2004—Aikhenvald 2004: 253)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Paumari (Aikhenvald, 2004—Aikhenvald 2004: 284)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Persian (Aikhenvald, 2004—Aikhenvald 2004: 38)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Portuguese (Aikhenvald, 2004—Aikhenvald 2004: 37)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Russian (Aikhenvald, 2004—Aikhenvald 2004)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Santali (Aikhenvald, 2004—Aikhenvald 2004: 131)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Spanish (Aikhenvald, 2004—Aikhenvald 2004)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Tuvaluan (Aikhenvald, 2004—Aikhenvald 2004: 359)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Ubykh (Aikhenvald, 2004—Aikhenvald 2004: 293)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Vlach Romani (Aikhenvald, 2004—Aikhenvald 2004: 38)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Wasco-Wishram (Aikhenvald, 2004—Aikhenvald 2004: 39)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Weyewa (Aikhenvald, 2004—Aikhenvald 2004: 359)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Zazaki (Aikhenvald, 2004—Aikhenvald 2004: 38-39)	(VSIAHQ)	0.00	0.38	0.00	1.00	1.05
Agul (Aikhenvald, 2004—Aikhenvald 2004: 143)	(VSIAH)(Q)	0.16	0.27	0.03	0.56	1.21
Chemehuevi (Aikhenvald, 2004—Aikhenvald 2004: 51)	(VSIAH)(Q)	0.16	0.27	0.03	0.56	1.21
Menomini (Aikhenvald, 2004—Aikhenvald 2004: 33)	(VSIAH)(Q)	0.16	0.27	0.03	0.56	1.21
Chipewyan (Aikhenvald, 2004—Aikhenvald 2004: 31)	(VSAQ)(IH)	0.21	0.27	0.06	0.36	1.30
Hare (Aikhenvald, 2004—Aikhenvald 2004: 31)	(VSAQ)(IH)	0.21	0.27	0.06	0.36	1.30
Kato (Aikhenvald, 2004—Aikhenvald 2004: 31)	(VSAQ)(IH)	0.21	0.27	0.06	0.36	1.30
Achagua (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Arabela (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Arrernte (Aikhenvald, 2004—Aikhenvald 2004: 33)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Baniwa (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Cashibo (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Cupeno (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Daw (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Enga (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Estonian (Aikhenvald, 2004—Aikhenvald 2004: 33)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Gaviao (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Guarani, Paraguayan (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Ignaciano (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Karitiana (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Kham (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Kombai (Aikhenvald, 2004—Aikhenvald 2004: 137)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Latvian (Aikhenvald, 2004—Aikhenvald 2004: 33)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Lezgian (Aikhenvald, 2004—Aikhenvald 2004: 31)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Livonian (Aikhenvald, 2004—Aikhenvald 2004: 33)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Omaha (Aikhenvald, 2004—Aikhenvald 2004: 33)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Pareci (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Piapoco (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Piro (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Potawatomi (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Resigaro (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Shoshone (Aikhenvald, 2004—Aikhenvald 2004: 33-34)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Sissala (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Surui (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Tauya (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Terena (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Warlpiri (Aikhenvald, 2004—Aikhenvald 2004: 62)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Waura (Aikhenvald, 2004—Aikhenvald 2004: 32)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Yankunytjatjara (Aikhenvald, 2004—Aikhenvald 2004: 33)	(VSIA)(HQ)	0.28	0.15	0.00	0.97	1.99
Comanche (Aikhenvald, 2004—Aikhenvald 2004: 50)	(VSIA)(H)(Q)	0.32	0.14	0.02	0.84	2.29
Dakota (Aikhenvald, 2004—Aikhenvald 2004: 51)	(VSIA)(H)(Q)	0.32	0.14	0.02	0.84	2.29
Tonkawa (Aikhenvald, 2004—Aikhenvald 2004: 51)	(VSIA)(H)(Q)	0.32	0.14	0.02	0.84	2.29
Abkhaz (Aikhenvald, 2004—Aikhenvald 2004: 29-30)	(VS)(IAHQ)	0.36	0.12	0.01	0.91	2.57
Albanian (Aikhenvald, 2004—Aikhenvald 2004: 40)	(VS)(IAHQ)	0.36	0.12	0.01	0.91	2.57
Arizona Tewa (Aikhenvald, 2004—Aikhenvald 2004: 42)	(VS)(IAHQ)	0.36	0.12	0.01	0.91	2.57
Armenian (Aikhenvald, 2004—Aikhenvald 2004: 39)	(VS)(IAHQ)	0.36	0.12	0.01	g.91	2.57
Bagvalal (Aikhenvald, 2004—Aikhenvald 2004: 155)	(VS)(IAHQ)	0.36	0.12	0.01	0.91	2.57
Balkan Slavic (Aikhenvald, 2004—Aikhenvald 2004: 40;158)	(VS)(IAHQ)	0.36	0.12	0.01	0.91	2.57
Bulgarian (Aikhenvald, 2004—Aikhenvald 2004: 288)	(VS)(IAHQ)	0.36	0.12	0.01	0.91	2.57
Cherokee (Aikhenvald, 2004—Aikhenvald 2004: 26-7)	(VS)(IAHQ)	0.36	0.12	0.01	0.91	2.57
Circassian (Aikhenvald, 2004—Aikhenvald 2004: 293)	(VS)(IAHQ)	0.36	0.12	0.01	0.91	2.57
Cree (Aikhenvald, 2004—Aikhenvald 2004: 42)	(VS)(IAHQ)	0.36	0.12	0.01	0.91	2.57

Cree/Montagnais/Naskapi (Aikhenvald, 2004—Aikhenvald 2004: 42)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Godoberi (Aikhenvald, 2004—Aikhenvald 2004: 28)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Hunzib (Aikhenvald, 2004—Aikhenvald 2004: 29)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Kalasha (Aikhenvald, 2004—Aikhenvald 2004: 28)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Kazakh (Aikhenvald, 2004—Aikhenvald 2004: 40)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Khowar (Aikhenvald, 2004—Aikhenvald 2004: 24)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Komi (Aikhenvald, 2004—Aikhenvald 2004: 28)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Kurdish (Aikhenvald, 2004—Aikhenvald 2004: 289)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Macedonian (Aikhenvald, 2004—Aikhenvald 2004: 40)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Mangap-Mbula (Aikhenvald, 2004—Aikhenvald 2004: 28)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Mari (Aikhenvald, 2004—Aikhenvald 2004: 28)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Megleno-Romanian (Aikhenvald, 2004—Aikhenvald 2004: 288)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Nepali (Aikhenvald, 2004—Aikhenvald 2004: 291)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Secoya (Aikhenvald, 2004—Aikhenvald 2004: 28)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Serbian (Aikhenvald, 2004—Aikhenvald 2004: 288)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Tajik (Aikhenvald, 2004—Aikhenvald 2004: 289)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Takelma (Aikhenvald, 2004—Aikhenvald 2004: 41)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Tarascan (Aikhenvald, 2004—Aikhenvald 2004: 41-42)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Tibetan (Aikhenvald, 2004—Aikhenvald 2004: 28)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Tsez (Aikhenvald, 2004—Aikhenvald 2004: 28)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Turkmen (Aikhenvald, 2004—Aikhenvald 2004: 40)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Udmurt (Aikhenvald, 2004—Aikhenvald 2004: 28)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Uyghur (Aikhenvald, 2004—Aikhenvald 2004: 40)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Uzbek (Aikhenvald, 2004—Aikhenvald 2004: 40)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Yanam (Aikhenvald, 2004—Aikhenvald 2004: 18)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Yukaghir (Aikhenvald, 2004—Aikhenvald 2004: 24;27)	(vs)(iahq)	0.36	0.12	0.01	0.91	2.57
Kwakiutl (Aikhenvald, 2004—Aikhenvald 2004: 59)	(vsi)(a)(h)(q)	0.41	0.11	0.02	0.83	3.40
Amdo Tibetan (Aikhenvald, 2004—Aikhenvald 2004: 45)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Archi (Aikhenvald, 2004—Aikhenvald 2004: 29)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Aymara (Aikhenvald, 2004—Aikhenvald 2004: 43)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Bora (Aikhenvald, 2004—Aikhenvald 2004: 44)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Capanawa (Aikhenvald, 2004—Aikhenvald 2004: 46)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Diyari (Aikhenvald, 2004—Aikhenvald 2004: 34)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Hopi (Aikhenvald, 2004—Aikhenvald 2004: 45)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Jarawara (Aikhenvald, 2004—Aikhenvald 2004: 23-4;26)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Koreguaje (Aikhenvald, 2004—Aikhenvald 2004: 44)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Maidu (Aikhenvald, 2004—Aikhenvald 2004: 46)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Moseten (Aikhenvald, 2004—Aikhenvald 2004: 44)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Ngiyambaa (Aikhenvald, 2004—Aikhenvald 2004: 34)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Ponca (Aikhenvald, 2004—Aikhenvald 2004: 33-34)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Quechua (Aikhenvald, 2004—Aikhenvald 2004: 43)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Retuara (Aikhenvald, 2004—Aikhenvald 2004: 49)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Sanuma (Aikhenvald, 2004—Aikhenvald 2004: 18;46)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Shasta (Aikhenvald, 2004—Aikhenvald 2004: 43)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Shilluk (Aikhenvald, 2004—Aikhenvald 2004: 43)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Skidegate Haida (Aikhenvald, 2004—Aikhenvald 2004: 46)	(vs)(ia)(hq)	0.42	0.08	0.00	1.00	8.79
Lillooet (Aikhenvald, 2004—Aikhenvald 2004: 59)	(vs)(i)(a)(hq)	0.44	0.08	0.00	0.96	9.54
Shipibo-Konibo (Aikhenvald, 2004—Aikhenvald 2004: 55)	(vs)(i)(a)(hq)	0.44	0.08	0.00	0.96	9.54
Thompson (Aikhenvald, 2004—Aikhenvald 2004: 59)	(vs)(i)(a)(hq)	0.44	0.08	0.00	0.96	9.54
Cora (Aikhenvald, 2004—Aikhenvald 2004: 57)	(vs)(ia)(h)(q)	0.47	0.08	0.00	0.92	9.86
Northern Embera (Aikhenvald, 2004—Aikhenvald 2004: 58)	(vs)(ia)(h)(q)	0.47	0.08	0.00	0.92	9.86
Southeastern Tepehuan (Aikhenvald, 2004—Aikhenvald 2004: 58)	(vs)(ia)(h)(q)	0.47	0.08	0.00	0.92	9.86
Mansi (Aikhenvald, 2004—Aikhenvald 2004: 31)	(v)(siahq)	1.00	0.26	0.23	0.75	14.70
Megrelian (Aikhenvald, 2004—Aikhenvald 2004: 31)	(v)(siahq)	1.00	0.26	0.23	0.75	14.70
Nenets (Aikhenvald, 2004—Aikhenvald 2004: 31)	(v)(siahq)	1.00	0.26	0.23	0.75	14.70
Northern Khanty (Aikhenvald, 2004—Aikhenvald 2004: 155-6)	(v)(siahq)	1.00	0.26	0.23	0.75	14.70
Svan (Aikhenvald, 2004—Aikhenvald 2004: 76)	(v)(siahq)	1.00	0.26	0.23	0.75	14.70
Turkish (Aikhenvald, 2004—Aikhenvald 2004: 155)	(v)(siahq)	1.00	0.26	0.23	0.75	14.70
Xakas (Aikhenvald, 2004—Aikhenvald 2004: 30)	(v)(siahq)	1.00	0.26	0.23	0.75	14.70
Meithei (Aikhenvald, 2004—Aikhenvald 2004: 31)	(viahq)(s)	1.00	0.27	0.25	0.75	14.70
Siona (Aikhenvald, 2004—Aikhenvald 2004: 46)	(v)(s)(iahq)	1.29	0.04	0.04	0.95	20.88
Barasano (Aikhenvald, 2004—Aikhenvald 2004: 51)	(v)(s)(ia)(hq)	1.35	0.01	0.00	0.97	24.12
Dulong (Aikhenvald, 2004—Aikhenvald 2004: 47)	(v)(s)(ia)(hq)	1.35	0.01	0.00	0.97	24.12
Eastern Pomo (Aikhenvald, 2004—Aikhenvald 2004: 52)	(v)(s)(ia)(hq)	1.35	0.01	0.00	0.97	24.12
Enets (Aikhenvald, 2004—Aikhenvald 2004: 47)	(v)(s)(ia)(hq)	1.35	0.01	0.00	0.97	24.12
Ladakhi (Aikhenvald, 2004—Aikhenvald 2004: 53)	(v)(s)(ia)(hq)	1.35	0.01	0.00	0.97	24.12
Macuna (Aikhenvald, 2004—Aikhenvald 2004: 52)	(v)(s)(ia)(hq)	1.35	0.01	0.00	0.97	24.12
Maricopa (Aikhenvald, 2004—Aikhenvald 2004: 47)	(v)(s)(ia)(hq)	1.35	0.01	0.00	0.97	24.12
Nganasan (Aikhenvald, 2004—Aikhenvald 2004: 47)	(v)(s)(ia)(hq)	1.35	0.01	0.00	0.97	24.12
Oksapmin (Aikhenvald, 2004—Aikhenvald 2004: 46)	(v)(s)(ia)(hq)	1.35	0.01	0.00	0.97	24.12
Qiang (Aikhenvald, 2004—Aikhenvald 2004: 44-45)	(v)(s)(ia)(hq)	1.35	0.01	0.00	0.97	24.12
Shibacha Lisu (Aikhenvald, 2004—Aikhenvald 2004: 54)	(v)(s)(ia)(hq)	1.35	0.01	0.00	0.97	24.12
Siriano (Aikhenvald, 2004—Aikhenvald 2004: 51-2)	(v)(s)(ia)(hq)	1.35	0.01	0.00	0.97	24.12
Tatuyo (Aikhenvald, 2004—Aikhenvald 2004: 51)	(v)(s)(ia)(hq)	1.35	0.01	0.00	0.97	24.12
Tucano (Aikhenvald, 2004—Aikhenvald 2004: 51-3)	(v)(s)(ia)(hq)	1.35	0.01	0.00	0.97	24.12
Washo (Aikhenvald, 2004—Aikhenvald 2004: 46)	(v)(s)(ia)(hq)	1.35	0.01	0.00	0.97	24.12
Desano (Aikhenvald, 2004—Aikhenvald 2004: 60)	(v)(s)(i)(a)(hq)	1.37	0.00	0.00	0.98	25.94
Hupa (Aikhenvald, 2004—Aikhenvald 2004: 31)	(v)(s)(i)(a)(hq)	1.37	0.00	0.00	0.98	25.94
Hupda (Aikhenvald, 2004—Aikhenvald 2004: 60)	(v)(s)(i)(a)(hq)	1.37	0.00	0.00	0.98	25.94
Kashaya (Aikhenvald, 2004—Aikhenvald 2004: 60)	(v)(s)(i)(a)(hq)	1.37	0.00	0.00	0.98	25.94
Mamainde (Aikhenvald, 2004—Aikhenvald 2004: 56)	(v)(s)(i)(a)(hq)	1.37	0.00	0.00	0.98	25.94
Pawnee (Aikhenvald, 2004—Aikhenvald 2004: 56)	(v)(s)(i)(a)(hq)	1.37	0.00	0.00	0.98	25.94
Southern Nambiquera (Aikhenvald, 2004—Aikhenvald 2004: 61)	(v)(s)(i)(a)(hq)	1.37	0.00	0.00	0.98	25.94
Tariana (Aikhenvald, 2004—Aikhenvald 2004: 1-3;52)	(v)(s)(i)(a)(hq)	1.37	0.00	0.00	0.98	25.94
Tsafiki (Aikhenvald, 2004—Aikhenvald 2004: 54)	(v)(s)(i)(a)(hq)	1.37	0.00	0.00	0.98	25.94
Tuyuca (Aikhenvald, 2004—Aikhenvald 2004: 60)	(v)(s)(i)(a)(hq)	1.37	0.00	0.00	0.98	25.94
Wichita (Aikhenvald, 2004—Aikhenvald 2004: 59-60)	(v)(s)(i)(a)(hq)	1.37	0.00	0.00	0.98	25.94
Wintu (Aikhenvald, 2004—Aikhenvald 2004: 60)	(v)(s)(i)(a)(hq)	1.37	0.00	0.00	0.98	25.94
Xamatauteri (Aikhenvald, 2004—Aikhenvald 2004: 18;56)	(v)(s)(i)(a)(hq)	1.37	0.00	0.00	0.98	25.94
Yuhup (Aikhenvald, 2004—Aikhenvald 2004: 292)	(v)(s)(i)(a)(hq)	1.37	0.00	0.00	0.98	25.94
Central Pomo (Aikhenvald, 2004—Aikhenvald 2004: 61)	(v)(s)(i)(a)(h)(q)	1.42	0.00	0.01	0.99	32.00
Fasu (Aikhenvald, 2004—Aikhenvald 2004: 62-3)	(v)(s)(i)(a)(h)(q)	1.42	0.00	0.01	0.99	32.00

Table S9: Attested evidentiality systems and their parameters according to the model. ᴠ-visual; ꜱ-sensory; ɪ-inferred; ᴀ-assumed; ʜ-hearsay; ǫ-quoted.