Appendix for Beyond Hill and Valley: A Sociogeographic Analysis of Argument Coding Complexity in the Eastern Himalayan Region

Appendix A

This appendix describes the sample, data collection and coding procedure in further detail.

1. Languages Sampled

Table 1 below details the sample. The sample was initially collected proportionally to size of subgroup, but after excluding languages for lack of adequate data sources, languages were included by convenience. However, this still results in a fairly genealogically diverse sample, as can be seen by the variety in subgroup and node below subgroup, given in the table below. The sample numbers to 32 languages. Data for K'man (glottocode:gema1234) and Asakian (glottocode:chak1270) were collected, but excluded for lack of adequate linguistic and ethnographic information to sufficiently code all variables with confidence.

Table 1: List of references for linguistic information

Language	Glottocode	Subgroup (node below top-level)	Node below subgroup	Citation
Rabha	rabh1238	Kochic	Bodo-Garo	Joseph 2007
Bodo-Mech	bodo1269	Bodo-Mech-Kachari	Bodo-Garo	Brahma 2015, Haokip 2018
Garo	garo1247	Bodo-Garo	Bodo-Garo	Burling 2003
Tiwa	tiwa1253	Sal (Brahmaputran)	Bodo-Garo	Dawson 2020a, 2020b
Turung	sing1264	Sal (Brahmaputran)	Jingpho-Asakian	Morey, 2010
Khiamniungan	khia1236	Sal (Brahmaputran)	Patkaian	Van Dam and Tham, p.c.
Hakhun Tangsa	noct1238	Sal (Brahmaputran)	Patkaian	Boro 2017
Hills Karbi	karb1241	Kuki-Chin-Naga	Karbic	Konnerth, 2014
Manipuri	mani1292	Kuki-Chin-Naga	Manipuri	Chelliah, 1997
Ao Naga	mong1332	Kuki-Chin-Naga	Angami-Ao	Coupe, 2007
Angami	anga1288	Kuki-Chin-Naga	Angami-Ao	Kevichusa-Ezung, 2014
Suansu	suan1234	Kuki-Chin-Naga	Tangkhul-Maring	Ivani, 2023
Zeme	zeme1240	Kuki-Chin-Naga	Zemeic	Chanu, 2017
Mizo	lush1249	Kuki-Chin-Naga	Kuki-Chin	Chhangte, 1993
Thado Chin	thad1238	Kuki-Chin-Naga	Kuki-Chin	Haokip, 2014
Lamkang	lamk1238	Kuki-Chin-Naga	Kuki-Chin	Thounaojam and Chelliah, 2007,

				Chelliah et al., 2019
Daai Chin	daai1236	Kuki-Chin-Naga	Kuki-Chin	So-Hartmann, 2009
Bori-Karko	bori1243	Macro-Tani	Eastern Tani	Lalrempuii, 2005
Milang	mila1245	Macro-Tani	Koro-Holon	Modi, 2017
Mising	miny1240	Macro-Tani	Mising-Padam- Miri-Minyong	Prasad, 1991
Galo	galo1242	Macro-Tani	Subansiri	Post, 2007
Kera'a	<u>idum1241</u>	Digarish	Digarish	Peck and Reinöhl, p.c.
Hrusso Aka	hrus1242	Hruso	Unclear	D'Souza, 2021
Tibetan	<u>tibe1272</u>	Bodic	Central Tibetan	DeLancey, 2003
Brokpa/Brokpake	brok1248	Bodic	Southern Tibetic	Wangdi, 2021
Bjokapakha	<u>bjok1234</u>	Bodic	Tshanglic	Grollmann, 2020
Tamangic	tama1367	Bodic	Ghale-Tamangic	Owen-Smith, 2015
Limbu	<u>limb1266</u>	Himalayish	Kiranti	van Driem, 1987
Lepcha	lepc1244	Himalayish	Himalayish	Plaisier, 2007
Puroik	puro1234	Kho-Bwa	Kho-Bwa	Lierberherr, 2017
Miji	miji1239	Miji	Miji	Weedall, 2021
Trung/Drung	<u>drun1238</u>	Nungish	Gunong	Perlin, 2020

2. Coding of sociocultural variables

The following table links ethnographic and other literature used to code sociocultural variables to the languages. Sometimes, linguistic sources (such as grammars) were re-used if they were the best available source on sociocultural information. Oftentimes, linguistic sources were used in conjunction with other sources to fully code sociocultural variables.

Table 2: List of references for sociocultural, agricultural and geographic information.

Language	Glottocode	Subgroup (node below top-level)	Node below subgroup	Citation
Rabha	<u>rabh1238</u>	Kochic	Bodo-Garo	Goswamy et al, 2024
Bodo-Mech	bodo1269	Bodo-Mech-Kachari	Bodo-Garo	Boro, 2024
Garo	garo1247	Bodo-Garo	Bodo-Garo	Burling, 1963
Tiwa	tiwa1253	Sal (Brahmaputran)	Bodo-Garo	Doloi et al, 2024
Turung	sing1264	Sal (Brahmaputran)	Jingpho-Asakian	Morey, 2010
Khiamniungan	khia1236	Sal (Brahmaputran)	Patkaian	Van Dam and Tham,
Killallilluligali	Killa1230	Sai (Biaiiiiapuuaii)	Falkaidii	p.c.
Hakhun Tangsa	noct1238	Sal (Brahmaputran)	Patkaian	Boro 2017
Hills Karbi	<u>karb1241</u>	Kuki-Chin-Naga	Karbic	Hansepi and Laisram, 2022
Manipuri	mani1292	Kuki-Chin-Naga	Manipuri	Meitei and Sharma, 2023
Ao Naga	mong1332	Kuki-Chin-Naga	Angami-Ao	Mills, 1926
Angami	anga1288	Kuki-Chin-Naga	Angami-Ao	Hutton, 1921
Suansu	suan1234	Kuki-Chin-Naga	Tangkhul-Maring	Singh and Choudhary, 2015
Zeme	zeme1240	Kuki-Chin-Naga	Zemeic	Longkumer, 2007

Mizo	lush1249	Kuki-Chin-Naga	Kuki-Chin	Chhangte, 1993
Thado Chin	thad1238	Kuki-Chin-Naga	Kuki-Chin	Lalthansangsa, 2018
Lamkang	lamk1238	Kuki-Chin-Naga	Kuki-Chin	Thounaojam and Chelliah, 2007
Daai Chin	daai1236	Kuki-Chin-Naga	Kuki-Chin	Chan and Takeda, 2016
Bori-Karko	bori1243	Macro-Tani	Eastern Tani	Raj, 2010
Milang	mila1245	Macro-Tani	Koro-Holon	Modi, 2017
Mising	miny1240	Macro-Tani	Mising-Padam- Miri-Minyong	Chutia, 2020
Galo	galo1242	Macro-Tani	Subansiri	Post, 2007
Kera'a	<u>idum1241</u>	Digarish	Digarish	Peck and Reinöhl, p.c.
Hrusso Aka	hrus1242	Hruso	Unclear	Nimachow et al. 2011
Tibetan	tibe1272	Bodic	Central Tibetan	DeLancey, 2003
Brokpa/Brokpake	brok1248	Bodic	Southern Tibetic	Wangdi, 2021
Bjokapakha	<u>bjok1234</u>	Bodic	Tshanglic	Akamatsu, 2015
Tamangic	tama1367	Bodic	Ghale-Tamangic	Owen-Smith, 2015
Limbu	<u>limb1266</u>	Himalayish	Kiranti	Limbu, 2013
Lepcha	lepc1244	Himalayish	Himalayish	Thapa and Allay, 2021
Puroik	puro1234	Kho-Bwa	Kho-Bwa	Ramjuk, 2018
Miji	miji1239	Miji	Miji	Weedall, 2021
Trung/Drung	<u>drun1238</u>	Nungish	Gunong	Perlin, 2020

Full bibliographic references for all data sources can be found at the end of this document. For languages where the citation says 'p.c.', the researcher reached out to academics known to be working on the language. They were asked to fill out a questionnaire, given below. The questionnaire also reflects my own methodology when coding information from a grammar or ethnographic data source.

3. Questionnaire

Lang: (glottocode), filled out by: (name)

- 1. Do markers on the noun distinguish core grammatical relations? (Are there case markers?) **If no, ignore Q2 -Q8**
- 2. Is the only argument of a one-place (intransitive) predicate marked? If so, by what forms? (affixes or adpositions)
- 3. Is an agentive argument of a one-place (intransitive) predicate marked? If so, by what forms? (affixes or adpositions)
- 4. Is a patientive argument of a one-place (intransitive) predicate marked? If so, by what forms? (affixes or adpositions)
- 5. Is the agentive argument of a two-place (transitive) predicate marked? If so, by what forms? (affixes or adpositions)
- 6. Is the patientive argument of a two-place (transitive) predicate marked? If so, by what forms? (affixes or adpositions)
- 7. Is the marker for the agentive argument of a transitive predicate optional? If so, under what conditions? (if known)
- 8. Is the marker for the patientive argument of a transitive predicate optional? If so, under what conditions? (if known)
- 9. Are there verbal indexes for core grammatical relations in affirmative declarative clauses? If yes, which macro-roles (of *intransitive subject*, *agent*, and *undergoer*) are indexed? If yes

and different in different TAME categories, answer for category with most distinctions. **If no, ignore Q10-Q18.**

- 10. Is the only argument of a one-place (intransitive) predicate indexed on the verb? If so, what persons and numbers can be indexed on the verb?
- 11. Is the agentive argument of a two-place (transitive) predicate indexed on the verb? If so, what persons and numbers?
- 12. Is the patientive argument of a two-place (transitive) predicate indexed on the verb? If so, what persons and numbers?
- 13. Are the indexes for intransitive argument the same as the ones for the agentive transitive argument?
- 14. Are the indexes for intransitive argument the same as the ones for the patientive transitive argument?
- 15. What is the maximal paradigm of agent and undergoer combinations for the transitive clause that cause some morphological marking on the verb?
- 16. Is there a hierarchy that determines the expression of verbal indexes?
- 17. Is there an inverse marker on the verb?
- 18. Are there any TAME requirements for the verbal indexes?
- 19. What is the pragmatically unmarked constituent order in the transitive clause?
- 20. Are other constituent orders in the transitive clause possible?
- 21. Any outstanding comments?

Some geographic/cultural questions, if possible:

- 1. What is the largest settlement where this language is primarily spoken?
- 2. How many speakers are there?
- 3. Is the language spoken as an L2 by any other speech communities?
- 4. What is the type of agriculture traditionally practiced? (intensive/irrigated, extensive/shifting/slash-and-burn/*jhum*, horticulture/gathering, or casual) (based on https://d-place.org/parameters/EA028#1/30/152)
- 5. What are/were the jurisdictional levels beyond the local community? (None/autonomous bands or villages, petty chiefdom, larger chiefdom, state, complex state) (based on https://d-place.org/parameters/EA033#1/30/153)
- 6. Any outstanding comments?

Appendix B

This appendix describes the statistical analyses in further detail.

1. Introduction

All code and results for the statistical analyses can be found in the R notebook *synthesis.rmd*, at https://github.com/patrickdas/ehr-argument-coding.

Table 3 below details the analyses conducted in this study. Some of these analyses which were not fully detailed in the paper for reasons of space are given in full below.

Table 3: Details of all statistical models used in this study.

Model type	Response	Predictor	Random	N
	variable	variable	effect	
Logistic regression	Hill/Valley	Elevation	None	31
Logistic regression	Hill/Valley	Terrain	None	31
		roughness		
Logistic regression	Hill/Valley	Elevation,	None	31

		Terrain		
		roughness		
Mixed effects	Complexity	Hill/Valley	Subfamily	31
Mixed effects	Complexity	Agricultural	Subfamily	31
		intensity		
Mixed effects	Complexity	L2 Status	Subfamily	31
Mixed effects	Complexity	Population	Subfamily	31
		Size		
Mixed effects	Complexity	Political	Subfamily	31
		Organization	_	

2. Hill/Valley classification as a predictor of complexity

The primary analysis of this study is whether Hill/Valley classification can effectively predict linguistic complexity. To this end, I ran six linear mixed effects models. Linear mixed-effects models were fit with hill_valley as a fixed effect (categorical: Hill, Split, Valley with Hill as reference) and a random effect for subfamily. The tables below present the estimated coefficients, standard errors, degrees of freedom, t-values, p-values, and R² values for each model with Limbu included. Models with Limbu excluded are available in the paper S4.2. Hill was set as the reference level for all models.

Table 4: Mixed-effects model predicting cell complexity from Hill/Valley classification with Limbu included. Model formula: scaled_cell_complexity \sim hill_valley + (1 | subfamily). N = 32 languages; 10 subfamilies

Term	Estimate (β)	Std. Error	df	t	p
Intercept (Hill)	0.14	0.25	9.79	0.58	0.58
Split	-0.55	0.42	25.32	-1.30	0.20
Valley	-0.10	0.471	28.79	-0.21	0.83
R^2 : Marginal = 0.051; Conditional = 0.106					

Table 5: Mixed-effects model predicting form complexity from Hill/Valley classification with Limbu included. Model formula: scaled_form_complexity \sim hill_valley + (1 | subfamily). N = 32 languages; 10 subfamilies

Term	Estimate (β)	Std. Error	df	t	p	
Intercept (Hill)	0.050	0.254	12.26	0.196	0.848	
Split	-0.329	0.426	26.30	-0.772	0.447	
Valley	0.056	0.478	28.86	0.117	0.907	
R^2 : Marginal = 0.022; Conditional = 0.083						

Table 6: Mixed-effects model predicting case complexity from Hill/Valley classification with Limbu included. Model formula: scaled_case_marking_complexity \sim hill_valley + (1 | subfamily). N = 32 languages; 10 subfamilies

Term	Estimate (β)	Std. Error	df	t	p
Intercept (Hill)	-0.264	0.239	12.09	-1.104	0.291
Split	0.542	0.414	26.57	1.309	0.202
Valley	0.712	0.462	28.85	1.541	0.134

3. Ecological analyses

Parameter

Three logistic regression models were fitted to test whether ecological variables predict Hill/Valley societal classification, using the sensitivity dataset (N = 31, excluding Limbu). The dependent variable was a binary classification where Hill = 1 and Valley/Split = 0.

Table 7: Logistic regression model predicting Hill/Valley classification from elevation and terrain roughness. Model formula: $hill_binary \sim scaled_altitude + scaled_stdev_slope$, family = binomial. N = 31 languages

Estimate Std. Error z-value n-value

	1 ul ullictel	Listinate	ota. Liioi	Z value	p value
	Intercept	0.341	0.369	0.922	0.356
	Elevation (scaled)	0.262	0.451	0.581	0.561
	Terrain roughness (scaled)	0.076	0.422	0.181	0.856
Model fit: AIC = 47.516, Null deviance = 42.165 (df = 30), Residual deviance = 41.516 (df = 28),					
	N = 31				

Table 8: Logistic regression model predicting Hill/Valley classification from elevation only. Model formula: hill_binary \sim scaled_altitude, family = binomial. N = 31 languages

Parameter	Estimate	Std. Error	z-value	e p-value
Intercept	0.341	0.369	0.924	0.355
Elevation (scaled)	0.301	0.396	0.760	0.448
Model fit: $AIC = 4$	5.548, Nul	l deviance =	42.165	($df = 30$), Residual deviance = 41.548 ($df = 29$),
N = 31				

Table 9: Logistic regression model predicting Hill/Valley classification from terrain roughness only. Model formula: hill_binary \sim scaled_stdev_slope, family = binomial. N = 31 languages

Parameter	Estimate	Std. Error	z-value	p-value
Intercept	0.329	0.366	0.898	0.369
Terrain roughness (scaled)	0.200	0.368	0.544	0.587
Model fit: AIC = 45.866, N	Iull devian	ce = 42.165	(df = 30)	, Residual deviance = 41.866 (df = 29),
N = 31				

3.1 Does coding Split as either Hill/Valley affect prediction by ecological variables?

In an alternate analysis, I checked whether coding Split languages as Hill instead affects the prediction by ecological variables. The results show that this reclassification does not improve ecological prediction. While the models show better overall fit (lower AIC values), the ecological predictors themselves remain non-significant across all models. Elevation and terrain roughness still fail to predict societal classification (all p > 0.4), reinforcing the conclusion that the ecological foundations of the Hill/Valley framework are not supported in this dataset.

Table 10: Logistic regression model predicting Hill/Valley classification from elevation and terrain roughness with alternative coding. Model formula: hill_valley_binary \sim scaled_altitude + scaled_stdev_slope, family = binomial. N = 31 languages.

Parameter	Estimate	Std. Error	z-value	p-value
Intercept	1.172	0.452	2.591	0.010**
Elevation (scaled)	0.327	0.610	0.536	0.592
Terrain roughness (scaled)	0.495	0.493	1.005	0.315
Model fit : AIC = 38.753, N	Iull devian	ce = 35.403	(df = 30)	, Residual deviance = 32.753 (df = 28),
N = 31				

Table 11: Logistic regression model predicting Hill/Valley classification from elevation only with alternative coding. Model formula: $hill_valley_binary \sim scaled_altitude$, family = binomial. N = 31 languages

Parameter	Estimate	Std. Error	z-value	p-value
Intercept	1.157	0.450	2.573	0.010*
Elevation (scaled)	0.617	0.551	1.120	0.263
Model fit : AIC = 3	7.847, Nul	l deviance =	35.403	(df = 30), Residual deviance = 33.847 $(df = 29)$,
N = 31				

Table 12: Logistic regression model predicting Hill/Valley classification from terrain roughness only with alternative coding. Model formula: $hill_valley_binary \sim scaled_stdev_slope$, family = binomial. N = 31 languages

Parameter	Estimate	Std. Error	z-value	p-value
Intercept	1.141	0.439	2.597	0.009**
Terrain roughness (scaled)	0.637	0.436	1.462	0.144
Model fit : AIC = 37.058, N	Iull devian	ce = 35.403	(df = 30)	, Residual deviance = 33.058 (df = 29),
N = 31				

4. Sociocultural analyses

4.1 Predicting complexity via L2_status

Linear mixed-effects models were fit with L2_status as a fixed effect (binary: 1 = present, 0 = not) and a random effect for subfamily. The tables below present the estimated coefficients, standard errors, degrees of freedom, t-values, p-values, and R^2 values for each model.

Table 13: Mixed-effects model predicting cell complexity from L2 status. Model formula: $scaled_cell_complexity \sim L2 + (1 \mid subfamily)N = 31 languages; 10 subfamilies$

Term	Estimate (β)	Std. Error	df	t	p		
Intercept	0.110	0.198	13.30	0.554	0.589		
L2	-0.618	0.274	28.22	-2.257	0.032*		
\mathbf{R}^2 : Marginal = 0.139; Conditional = 0.222							

Table 14: Mixed-effects model predicting form complexity from L2 status. Model formula: $scaled_form_complexity \sim L2 + (1 \mid subfamily)N = 31 languages; 10 subfamilies$

Term	Estimate (β)	Std. Error	df	t	p	
Intercept	-0.008	0.193	16.30	-0.043	0.966	
L2	-0.434	0.261	28.40	-1.663	0.107	
R^2 : Marginal = 0.079; Conditional = 0.189						

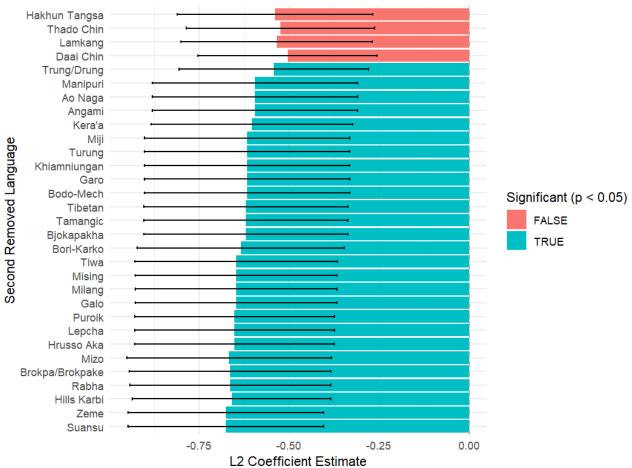
Table 15: Mixed-effects model predicting case-marking complexity from L2 status. Model formula: $scaled_case_marking_complexity \sim L2 + (1 \mid subfamily)N = 31 languages; 10 subfamilies$

Term	Estimate (β)	Std. Error	df	t	p	
Intercept	-0.233	0.251	8.54	-0.929	0.378	
L2	0.556	0.360	27.81	1.546	0.133	
R^2 : Marginal = 0.072; Conditional = 0.130						

4.2 Testing robustness of effect of L2_status on cell complexity

To test whether the significant relationship found between L2_status and cell complexity was truly robust, I wrote a function to loop through the revised dataset (excluding Limbu) to exclude one language at a time, to see whether the relationship maintained. The code can be found in the associated R-markdown document (*synthesis-code*) on the github, while I provide a graphical representation of the results of this process here.

L2 Effect on Cell Complexity - Drop-One Analysis Effect of removing each language after Limbu already removed



The median p-value is 0.038, while the coefficient estimates range from -0.676 to -0.504. The result is significant in 27 out of 31 iterations.

4.3 Predicting complexity via agricultural intensity

Linear mixed-effects models were fit with **agricultural intensity** as a fixed effect (binary: $1 = \frac{1}{2}$ Intensive/Irrigated, $0 = \frac{1}{2}$ Other) and a random effect for subfamily. The tables below present the estimated coefficients, standard errors, degrees of freedom, t-values, p-values, and R^2 values for each model.

Table 16: Mixed-effects model predicting cell complexity from agricultural intensity. Model formula: $scaled_cell_complexity \sim agriculture_binary + (1 | subfamily)N = 31 languages; 10 subfamilies$

Term	Estimate (β)	Std. Error	df	t	p
Intercept	-0.054	0.185	8.09	-0.292	0.777
Agricultural intensity	-0.533	0.381	27.24	-1.398	0.173
R^2 : Marginal = 0.057;	Conditional =	0.158			

Table 17: Mixed-effects model predicting form complexity from agricultural intensity. Model formula: $scaled_form_complexity \sim agriculture_binary + (1 | subfamily)N = 31 languages; 10 subfamilies$

Term	Estimate (β)	Std. Error	df	t	p
Intercept	-0.104	0.177	10.94	-0.588	0.568
Agricultural intensity	-0.506	0.350	27.72	-1.446	0.159
R^2 : Marginal = 0.059;	Conditional =	0.185			

Table 18: Mixed-effects model predicting case-marking complexity from agricultural intensity. Model formula: $scaled_case_marking_complexity \sim agriculture_binary + (1 | subfamily)N = 31 languages; 10 subfamilies$

Term	Estimate (β)	Std. Error	df	t	p
Intercept	-0.046	0.207	5.39	-0.222	0.832
Agricultural intensity	+0.247	0.501	27.21	+0.494	0.626
R^2 : Marginal = 0.008; Conditional = 0.022					

4.4 Predicting complexity via speaker population

Linear mixed-effects models were fit with **scaled_population** as a fixed effect (binary: $1 = \frac{1}{2}$ Intensive/Irrigated, $0 = \frac{1}{2}$ Other) and a random effect for subfamily. The tables below present the estimated coefficients, standard errors, degrees of freedom, t-values, p-values, and R^2 values for each model.

Table 19: Mixed-effects model predicting cell complexity from speaker population size. Model formula: $scaled_cell_complexity \sim scaled_population + (1 | subfamily)N = 31 languages; 10 subfamilies$

Term	Estimate (β)	Std. Error	df	t	p
Intercept	-0.054	0.185	8.09	-0.292	0.777
Agricultural intensity	-0.533	0.381	27.24	-1.398	0.173
\mathbf{R}^2 : Marginal = 0.027; Conditional = 0.134					

Table 20: Mixed-effects model predicting form complexity from speaker population size. Model formula: scaled_form_complexity \sim scaled_population + $(1 \mid subfamily)N = 31$ languages; 10 subfamilies

Term	Estimate (β)	Std. Error	df	t	p
Intercept	-0.104	0.177	10.94	-0.588	0.568
Agricultural intensity	-0.506	0.350	27.72	-1.446	0.159
R^2 : Marginal = 0.009;	Conditional =	0.114			

Table 21: Mixed-effects model predicting case-marking complexity from speaker population size. Model formula: $scaled_case_marking_complexity \sim scaled_population + (1 | subfamily)N = 31 languages; 10 subfamilies$

Term	Estimate (β)	Std. Error	df	t	p
Intercept	-0.046	0.207	5.39	-0.222	0.832
Agricultural intensity	0.247	0.501	27.21	0.494	0.626
\mathbf{R}^2 : Marginal = 0.014; Conditional = 0.042					

4.5 Predicting complexity via political organization

Linear mixed-effects models were fit with **political_organization** as a fixed effect (binary: 1 = State, 0 = Acephalous/other) and a random effect for subfamily. The tables below present the estimated coefficients, standard errors, degrees of freedom, t-values, p-values, and R^2 values for each model.

Table 22: Mixed-effects model predicting cell complexity from political organization. Model formula: scaled_cell_complexity \sim political_organization + (1 | subfamily). N = 31 languages; 10 subfamilies

Term	Estimate (β)	Std. Error	df	t	p
Intercept	-0.075	0.187	7.69	-0.403	0.698
Political organization	-0.492	0.428	28.47	-1.149	0.260
\mathbf{R}^2 : Marginal = 0.045; Conditional = 0.176					

Table 23: Mixed-effects model predicting form complexity from political organization. Model formula: scaled_form_complexity \sim political_organization + $(1 \mid subfamily)N = 31$ languages; 10 subfamilies

Term	Estimate (β)	Std. Error	df	t	p
Intercept	-0.122	0.178	10.46	-0.685	0.508
Political organization	-0.482	0.393	28.57	-1.226	0.230
\mathbf{R}^2 : Marginal = 0.045; Conditional = 0.176					

Table 24: Mixed-effects model predicting case-marking complexity from political organization. Model formula: scaled_case_marking_complexity \sim political_organization + $(1 \mid subfamily)N = 31$ languages; 10 subfamilies

Term	Estimate (β)	Std. Error	df	t	p
Intercept	-0.013	0.204	4.38	-0.062	0.953
Political organization	0.049	0.553	28.90	0.088	0.930
\mathbf{R}^2 : Marginal = 0.000; Conditional = 0.031					

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