Supplementary Information for Cultural transmission, networks, and clusters among Austronesian-speaking peoples

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154 S1 Supplementary Results

S1.1 Imputation

- 156 **Iterative imputer.** Iterative imputation works from least to most missing values per column using the imputed values in successive iterations. At each iteration, one feature (cultural trait) is chosen
- 158 as the dependent variable *y*, and the other features are treated as independent variables, *X*. Then features *y* are iterated according to which of them has the least missing values, assuming the inferred
 160 values from the previous iteration. A regression model is fitted to the (*X*, *y*) pairs without missing
- values. The fitted regression model is then used to predict missing y values, ending the current 162 iteration. At the next iteration, another feature is chosen as y, a new regression model is fitted and
- used to predict missing values, and so on, until all missing features are assigned an expected value.
- 164 We used the IterativeImputer implementation in *Scikit-learn* (Pedregosa et al., 2011), a Python machine learning library. For the initialization step, where a first estimate is obtained, missing values
- 166 were replaced by the column mean. The following regression models were considered: Bayesian ridge regression (IIBR), nearest neighbors regression with five or ten neighbors (IIKNN5, IIKNN10),
- 168 decision tree regression (IIDT), and random forest regression (IIRF).

Comparison of imputation strategies. There was a strong correlation between MSE (mean squared
 error) and accuracy across all datasets and imputation approaches (see Figure S23). Thus, we focused on differences in effect size imposed by differing MSEs and generally found small differences in the

- 172 accuracy of VBPCA imputation relative to other imputation algorithms across the 50 replicates and the four Austronesian datasets. None of the imputers consistently outperformed VBPCA. Across datasets
- IIBR was 3.9% less accurate (95% CI 1.1 3.8%, *Religion*) and 2.1% (95% CI 1.5 2.8%; *Social Organization*) more accurate than VBPCA. IIDT was between 11.8% (9.5-14.1%, *Cultural Interaction*)
- less accurate and 0.5% (0.01–1.0%) more accurate than VBPCA. IIKNN5 between 3.3% (0.5–6.1%, *Cultural Interaction*) less and 2.3% (0.8–3.8%, *Religion*) more accurate than VBPCA. IIKNN10
- 178 between 4.7% (2.2–7.1%, *Cultural Interaction*) less and 4.0% (2.6–5.4%, *Religion*) more accurate. IIRF between 11.5% (9.1-13.9%, *Cultural Interaction*) less and 1.4% (0.9-1.9%, *Social Organization*)
- 180 more accurate. Simple (replacement of missing values by column means) between 3.4% (2.9-3.8%, *Social Organization*) less and 2.3% (1.0-3.7%, *Religion*) more accurate (see Figure S21).

182 S1.2 Denoising

The proportion of cumulative variance explained by the optimal number of principal components
184 for data reconstruction as determined by VBPCA using classical PCA after filling missing values via different imputation approaches is a measure of the degree of noise VBPCA filters for each dataset. As

- 186 can be seen in Figure S22, the *Social Organization* dataset has the most noise filtered out, followed by *Religion, Subsistence*, and finally by *Cultural Interaction*. The degree of variance explained by each
- 188 of the considered imputation strategies is tightly clustered by the dataset and the proportion of noise filtered is not solely a function of the number of significant components, the size of the original data,
- 190 or the number of PCs for the optimal reconstruction (see Figure S22, and Anonymous, 2023).

S1.3 Effect of missing data on downstream analysis

- 192 We checked for differences in the degree of missingness between linguistic outlier and non-outlier cultures in each cultural class and correlations between how much data was missing and archetype
- 194 loadings. We found no significant differences in the proportion of missing data, except for the Agriculture archetype, between outliers and non-outliers in any of the cultural classes using Dunn's test
- 196 for stochastic dominance with null hypotheses that there was no difference in stochastic dominance (Table S11).
- 198 We also checked for correlation between extreme (max and min) archetype values for each culture and each cultural class and did not find significant correlations between extreme values and the degree of
- 200 missingness (Table S12).

S1.4 Naming of archetypes

- 202 **Social Organization.** Analysis of the *Social organization* class of variables revealed five cultural archetypes, which we name Duolateral Communities, Patrilineal Clans, Neo-local Kindreds, Matri-
- 204 lineal Clans, and Ambilineal Chiefdoms, based on their correlation with the original variables in this class (Figure S12). Ambilineality is associated with chiefdoms because cultures belonging to this
- 206 archetype were also associated with *social stratification into a hereditary aristocracy and a lower* class (Spearman correlation, $\rho = 0.45$, p < 0.001), which is a classic characteristic of chiefdoms
- 208 (Earle, 1997; Feinman and Marcus, 1998; Yoffee, 2005). Duolateral indicates the presence of both matrilineal and patrilineal kin groups; bilateral indicates the absence of matri/patrilineal kin groups
- 210 and the presence of cognatic kin groups; ambilineal means kin reckoning of both matrilineal and patrilineal kin without set order (Kirby et al., 2016; Fortunato, 2019).
- 212 **Subsistence.** We found four cultural archetypes for the *Subsistence* cultural class, which we name Roots & Pigs, Grains & Cattle, and Fish & Fruits (Figure S14). In addition to modes of subsistence
- 214 for which the archetypes are named, there were significant differences in labor practices among archetypes. Grains & Cattle archetype is associated with the importance of weaving, pottery-making,
- 216 and metal-working: this archetype is positively associated with *weaving by most adult females* ($\rho = 0.66, p < 0.001$); *metal-working as a craft specialty* ($\rho = 0.74, p < 0.001$), which is *male-dominated*
- 218 ($\rho = 0.74, p < 0.001$); and *pottery-making by most adults* ($\rho = 0.61, p < 0.001$). The Grain & Cattle is also notable for being the only archetype with a *presence of slavery in the past* ($\rho = 0.46, p < 0.001$).
- In contrast, the Roots & Pigs and Fish & Fruit archetypes are both associated with the *lack of importance* of metal-working ($\rho = 0.55$, p < 0.001 for both); the Fish & Fruit archetype is associated with the *lack*
- 222 of importance of pottery-making ($\rho = 0.43, 0.42$ respectively p < 0.001 for both) and the Roots & PIgs and Fish & Fruit are associated with having never practiced slavery ($\rho = 0.38, 0.39, p < 0.001$
- for both). Finally, alone among the archetypes, Agriculture is associated with a large degree of missing data and the majority of subsistence coming from agriculture of unspecified type.
- 226 **Religion.** We found four archetypes for the *Religion* class, which we name Heroes, *Mana*, Autonomous Communities, and Animism (Figure S11). The Heroes archetype is also associated with the *absence*
- 228 of god(s) ($\rho = 0.35, p < 0.001$), combined political & religious authority ($\rho = 0.47, p < 0.001$), a

religious community composed of the entire society ($\rho = 0.4, p < 0.001$). Mana is associated with

- 230 super-local (multiple-community) religious authority ($\rho = 0.51, p < 0.001$), and social hierarchy tapu (a tapu is a spritual restriction; $\rho = 0.37, p < 0.001$). Finally, the Autonomous Communities
- archetype is associated with the *absence of religious authority* ($\rho = 0.53$, p < 0.001) and *piercing as a rite* ($\rho = 0.43$, p < 0.001).
- 234 **Cultural Interaction.** Two archetypes, which we name Isolation and Contact, were found in the *Cultural Interaction* cultural class (Figure S16). Because there are only two archetypes, the correlations
- 236 for isolation and contact are opposite. Contact is positively associated with *the presence of pre-Austronesian populations* ($\rho = 0.85$, p < 0.001), *occasional conflict within* and *between communities*
- 238 ($\rho = 0.45, p < 0.001$ for both), and *frequent* or *high levels of warfare with other societies* ($\rho = 0.67, 0.52, p < 0.001$ for both). Isolation is positively associated with *occasional contact* ($\rho = 0.67, 0.52, p < 0.001$ for both).
- 240 0.49, p < 0.001), both *frequent* and *moderate conflict within communities* ($\rho = 0.73, 0.46, p < 0.001$ for both), and *rare warfare with other societies* ($\rho = 0.82, p < 0.001$).

242 S1.5 Cultural cluster analysis

Subsistence. In some circumstances the raw materials needed to produce a good may need to be
locally available. However, the lack of local resources to produce an item does not necessarily indicate its absence in a society. The Roots & Pigs and Fish & Fruit archetypes are associated with *lack of*

- 246 *importance of metalworking*; the Fish & Fruit archetypes are associated with *lack of importance of pottery-making*. This lack of manufacture does not indicate a lack of use. Large amounts of pottery
- 248 exist in the archaeological records of islands associated with the Fish & Fruit and Roots & Pigs archetypes (Spriggs, 1995). Genetic variation in "canoe plants" and "canoe animals" the corpus
- 250 of plants and animals taken in canoes along the path of migration together with occasional contact even for cultures belonging to the *Isolation* archetype suggests the presence of trade networks between
- islands (Whistler, 2009; Wilme et al., 2016). Thus, cultures belonging to the Fish & Fruit, and Roots& Pigs archetypes may have traded locally produced products for pots and metal goods.
- **Cultural Interaction.** Isolation is most common in Micronesia, Outer Melanesia, and Polynesia, likely due to the greater distances between neighbors in these regions (Micronesia 236.5 ± 46.1 km,
- 256 Outer Melanesia 56.5 \pm 15.7 km, Polynesia 747.7 \pm 220.7 km (mean \pm standard error); Figure S1-S3, and Table S1). Beyond *frequency of interaction with other societies*, the Contact archetype is
- 258 positively associated with the presence of pre-Austronesian populations, occasional conflict within and between communities, and frequent or high levels of warfare with other societies. The results
- 260 further indicate that variation in degree of isolation is primarily driven by distance from neighbors and the mainland.
- 262 Isolation is positively associated with both *frequent* and *moderate conflict within communities*, and *rare warfare with other societies*. These results suggest outwardly and inwardly driven conflict,
- 264 respectively: the primary source of conflict in societies belonging to the Isolation archetype is at the inter- and intra- community levels, whereas societies in the contact archetype appear more focused on
- 266 conflict with neighbors. Potential explanations for this difference include opportunity (closer proximity

to neighbors, see Figures S7 to S9, Table S1 and González-Ruibal, 2019) and the presence of near 268 neighbors acting as an internally unifying force and target for conflict (González-Ruibal, 2019).

S1.6 Hierarchical clustering

- 270 We performed agglomerative hierarchical clustering on the archetypes combined with non-cultural features such as island type, size, and elevation to identify clusters across cultural classes, geography,
- 272 and ecology (Figure S18). This analysis reveals three clusters for the 65 cultures that are common between EA and Pulotu. Notably, two of these clusters correspond well to the "Resource Competi-
- 274 tion" and "Resource Defense" clusters described by (Karin and Alon, 2018, Figure S18). Resource competition correlates with Coral Atolls, *Mana*, Isolation, and Fish & Fruit, and geographically corre-
- 276 sponds to the Polynesian islands. Resource defense is geographically associated with Melanesia, and is correlated with Contact, larger islands, and Grain & Cattle (Figure S18). These findings indicate
- 278 that the "ecotypes" suggested by Karin and Alon (2018) are robust to inference method. However, our results also indicate that care must be taken when collapsing cultural data into fewer features. In
- 280 effect performing hierarchical clustering on the combined archetypes amounts to iterative dimension reduction. Such approaches may remove meaningful cultural variation (for example finding three
- 282 hierarchical clusters on the combined data versus finding a total of 15 archetypes across four cultural classes).

284 S2 Supplementary tables and figures

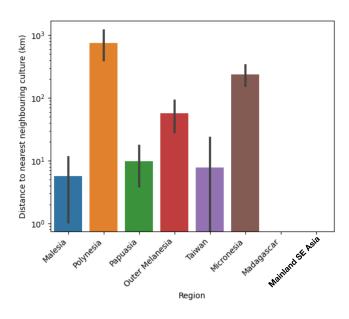


Figure S1: Regional distance comparison to nearest neighboring culture. A value of 0 indicates that all regional cultures have an immediate neighbor.

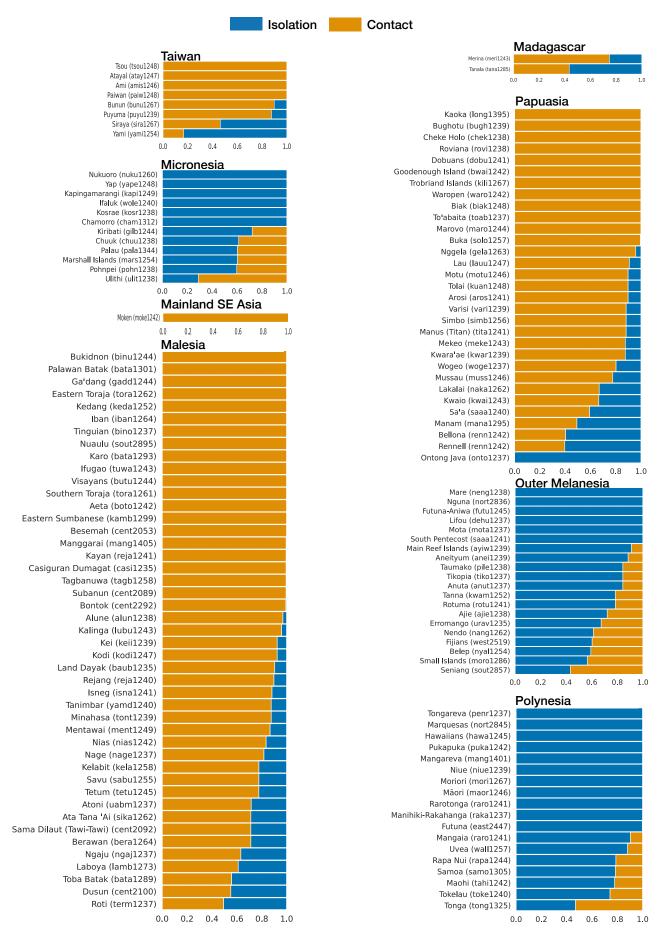


Figure S2: Cultural Interaction bar-plots with respect to archetypes. Cultures are separated by geographic region.

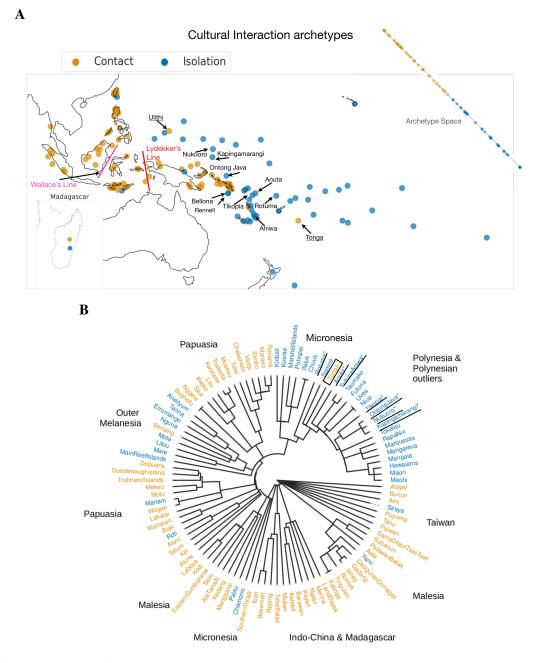


Figure S3: Variation in Cultural Interaction. Over (**a**) geography, (**inset**) archetypes (the Polynesian outliers in this data are labeled with text arrows), and (**b**) Austronesian language tree. Lydekker's line and Wallace's line are proposed geographic barriers separating the Asian and Australian biospheres (see Ali and Heaney, 2021).

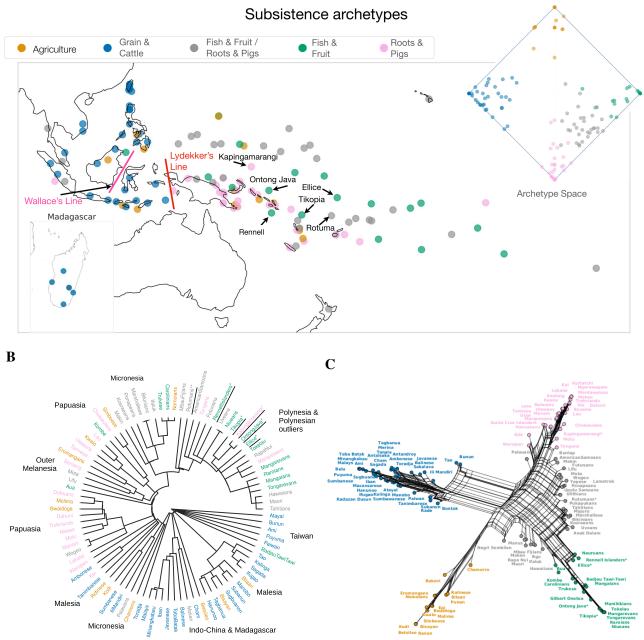


Figure S4: Variation in Subsistence. Austronesian cultures, colored by most likely cluster, over (**a**) geography, (**inset**) archetypes (the Polynesian outliers in this data are labeled with text arrows), and (**b**) Austronesian language tree. Lydekker's line and Wallace's line are proposed geographic barriers separating the Asian and Australian biospheres (see Ali and Heaney, 2021).

A

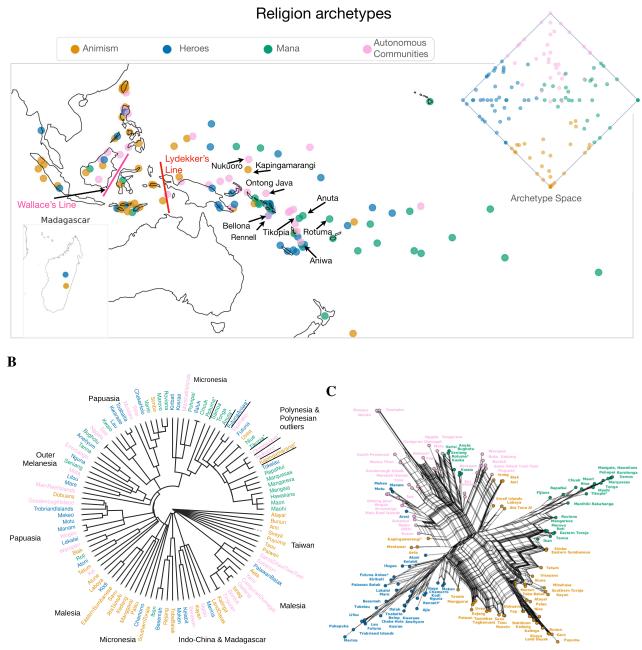


Figure S5: Variation in Religion. Austronesian cultures, colored by most likely cluster, over (**a**) geography, (**inset**) archetypes (the Polynesian outliers in this data are labeled with text arrows), and (**b**) Austronesian language tree. Lydekker's line and Wallace's line are proposed geographic barriers separating the Asian and Australian biospheres (see Ali and Heaney, 2021).

A

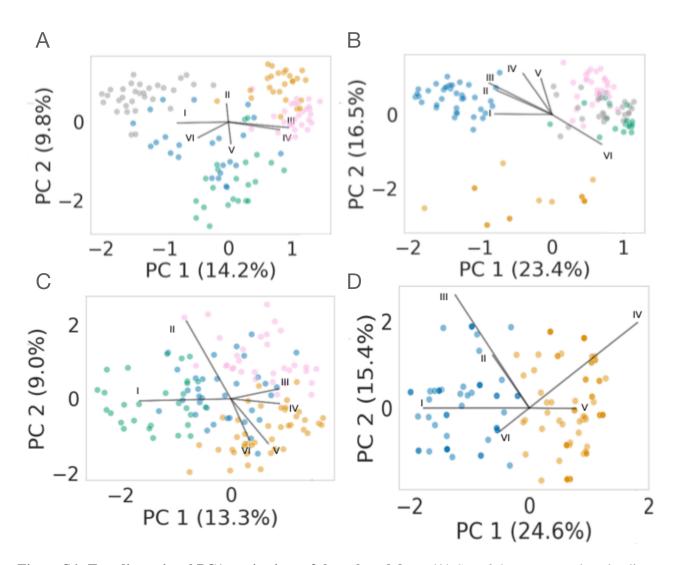


Figure S6: Two-dimensional PCA projections of the cultural data. (A) Social Organization data; loadings are: I) Token bride price, II) Polygamy, III) Significant bride price prevailing pattern, IV) Bride service, V) Domestic organization: small extended families, VI) Significant bride price, alternative pattern. Thus, PC 1 captures differences in domestic organization and PC 2 differences in transactions/exchange at marriage. (B) Subsistence data; loadings are I) Animal husbandry 46-55% of subsistence, II) Fishing 16-25% of subsistence, III) Fishing 6-15% of subsistence, IV) Agriculture 0-5% of subsistence, V) Complete absence of agriculture, VI) Fishing 36-45% of subsistence. Thus, PC 1 captures differences in the importance of animal husbandry and fishing and PC 2 differences in the importance of agriculture. (C) Religion data; loadings are I)Actions of others after one has died do not affect the nature of one's afterlife, II) No myth of man's creation, III) Belief in cultural heroes is a major focus, IV) Belief in nature spirit(s) a major focus, V) Creationist myth of humans, VI) Actions while living are the principal determinant of the nature of one's afterlife. Thus, PC 1 captures differences in religious focus, and PC 2 differences in creation myths. (D) Cultural Interaction data; loadings marked by I) Pre-Austronesians absent, II) Common conflict between communities, III) Occasional conflict between communities, IV) Frequent conflict with other societies, V) Pre-Austronesians present, VI) Common conflict with other societies. Thus, PC 1 captures the presence/absence of pre-existing populations at the time of settlement, and PC 2 captures differences in the nature and frequency of conflict within and between societies.

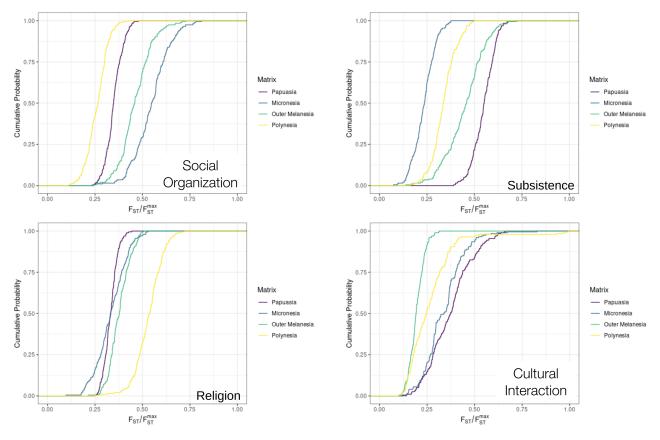


Figure S7: Within geographic region variability of archetype sets. Cumulative density plots for F_{st}/F_{st}^{max} , which is proportional to Dirichlet distribution variance in Polynesia and regions containing Polynesian outliers.



Figure S8: Cultural Interaction network. Due to the binary nature of these archetypes, the network reduces to a tree with only two branches.

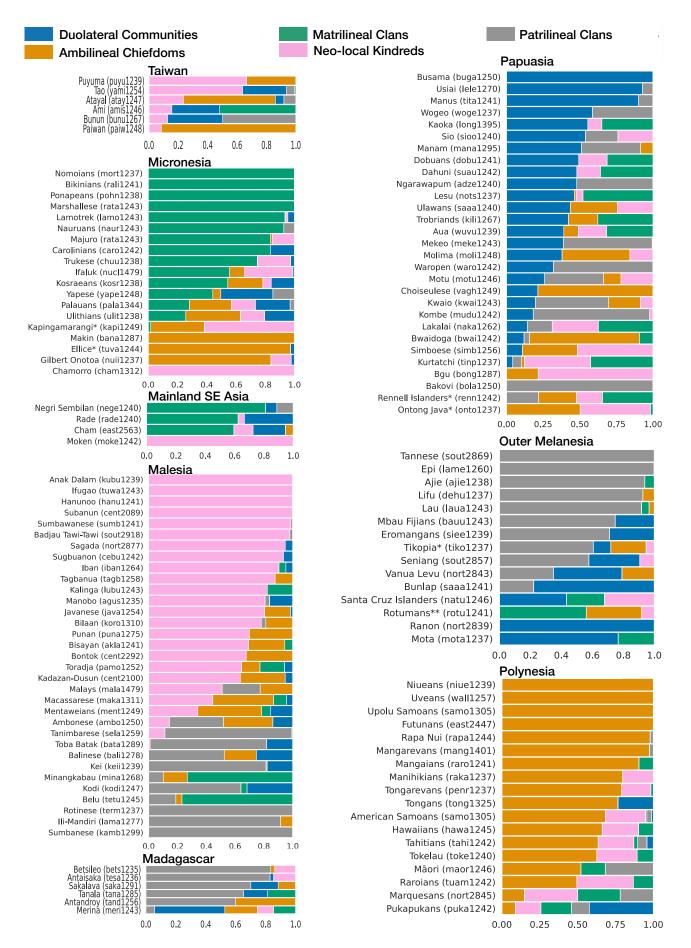


Figure S9: Social Organization bar-plots with respect to archetypes. Cultures are separated by geographic region.

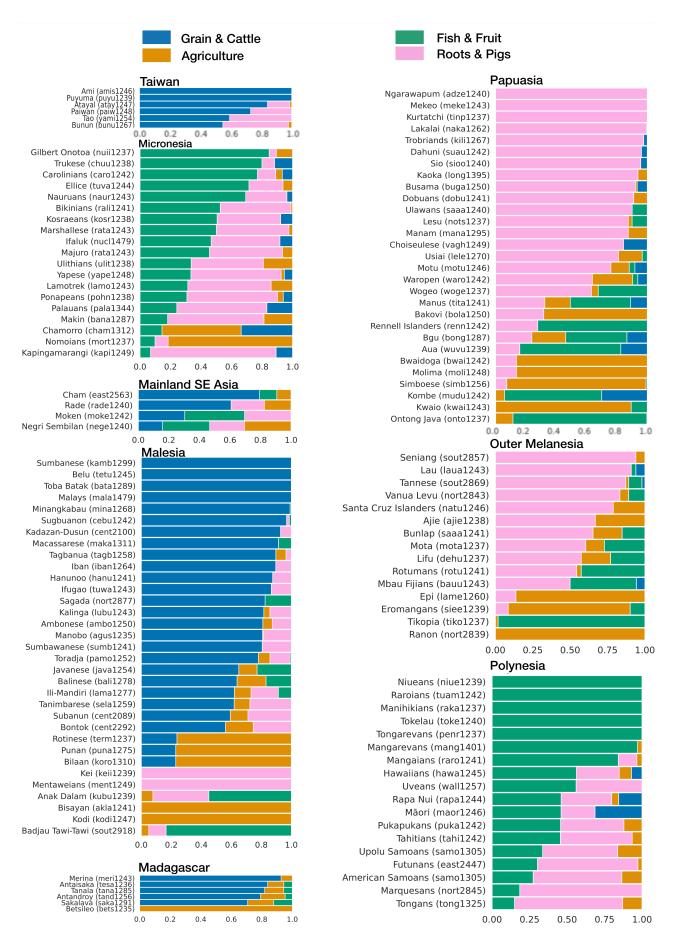


Figure S10: Subsistence bar-plots with respect to archetypes. Cultures are separated by geographic region.

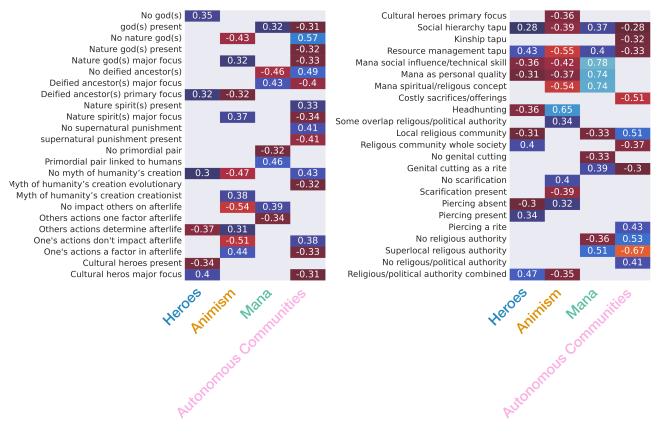


Figure S11: Binarized Religion trait correlations with archetypes. Only features (n = 89) that have correlation P-values less than 0.001 after FDR correction are displayed (n = 47).

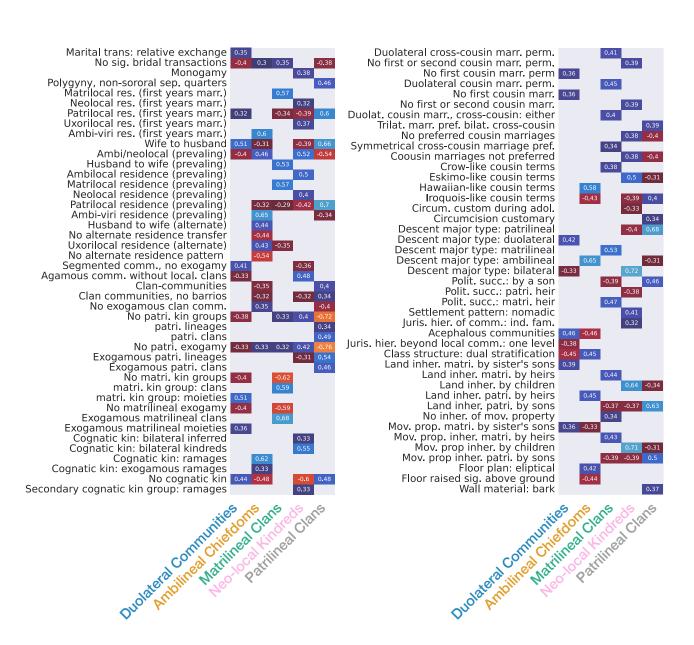


Figure S12: Binarized Social Organization trait correlations with archetypes. Only features (n = 236) that have correlation P-values less than 0.001 after FDR correction are displayed (n = 87).

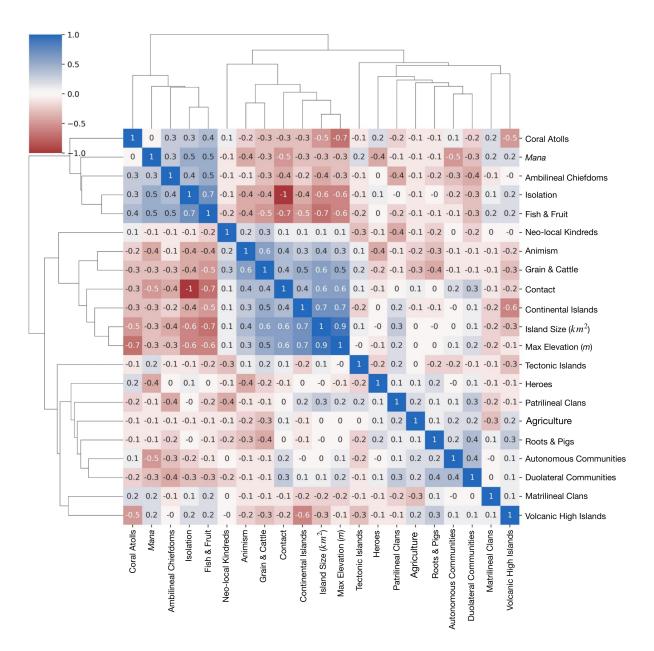


Figure S13: Hierarchical cluster-map on the Spearman correlation matrix for both archetypes and non-cultural features across the 65 cultures in common between EA and Pulotu. There are three clusters that appear to correspond to the results of Karin and Alon (2018).

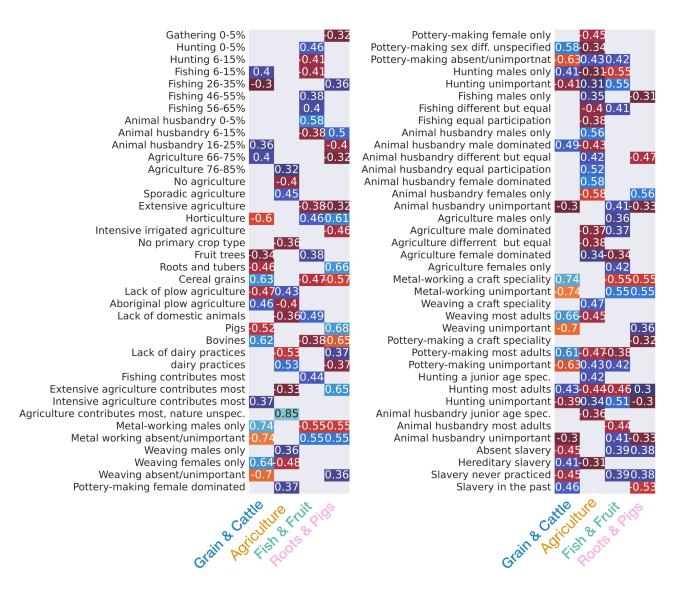


Figure S14: Binarized Subsistence trait correlations with archetypes. Only features (n = 121) that have correlation P-values less than 0.001 after FDR correction are displayed (n = 76). Note that the Agriculture archetype is missing many features on average (Table S11) and is primarily associated with (1) this missigness and (2) agriculture of unspecified type providing most of subsistence.

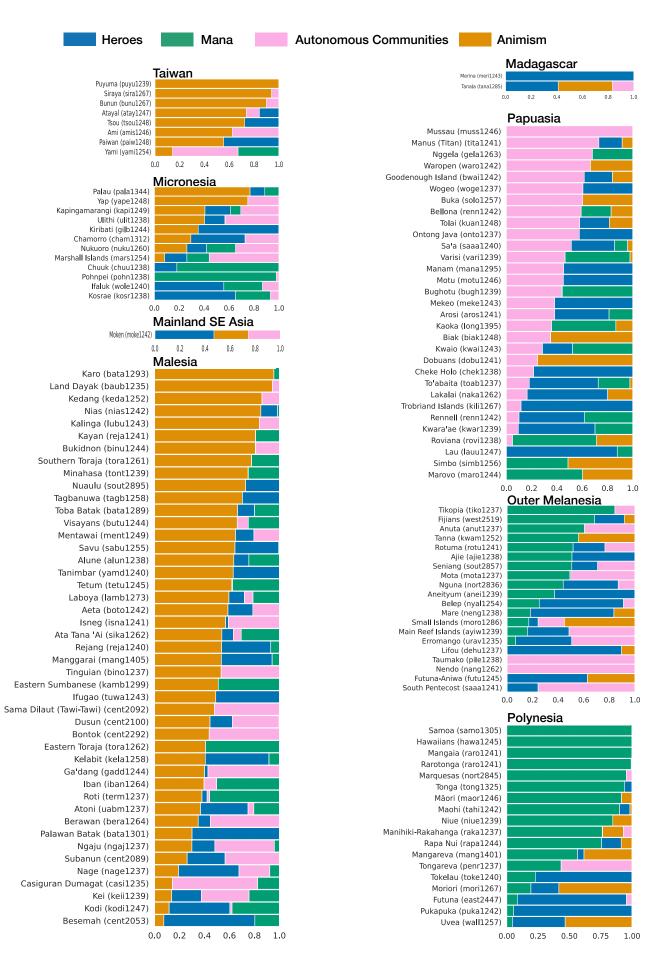


Figure S15: Religion bar-plots with respect to archetypes. Cultures are separated by geographic region.

Pre-Austronesians present	-0.85	0.85
Frequent contact	-0.88	88.0
Occasional contact	0.49	-0.49
Endemic conflict within communities	0.73	-0.73
Moderate conflict within communities	0.46	-0.46
Occasional conflict within communities	-0.46	0.46
Occasional conflict between communities	-0.45	0.45
Endemic warfare	-0.67	0.67
Frequent warfare	-0.51	0.51
Rare warfare	0.82	-0.82
	Isolation	Contact

Figure S16: Binarized Cultural Interaction trait correlations with archetypes. Only features (n = 18) that have correlation P-values less than 0.001 after FDR correction are displayed (n = 10).

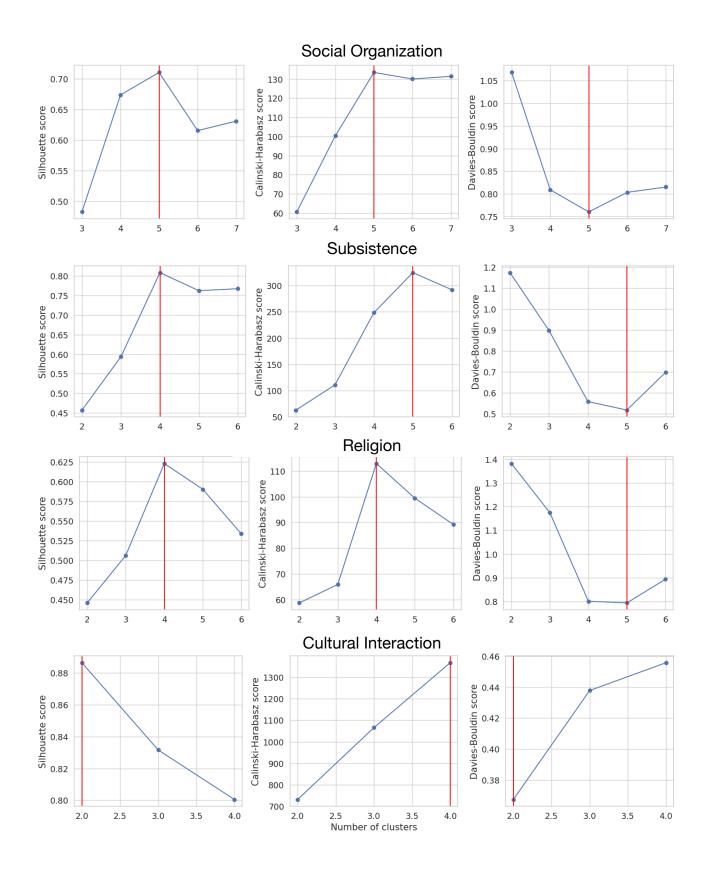


Figure S17: Kmeans clustering metrics for distances calculated in archetype space. Optimal numbers of Kmeans clusters for each data class (rows) according to the silhouette score (column 1), the variance ratio criterion (column 2), and the Davies-Bouldin score (column 3). For all archetypes except *Subsistence*, the optimal number of clusters on the archetypes is the same as the number of archetypes.

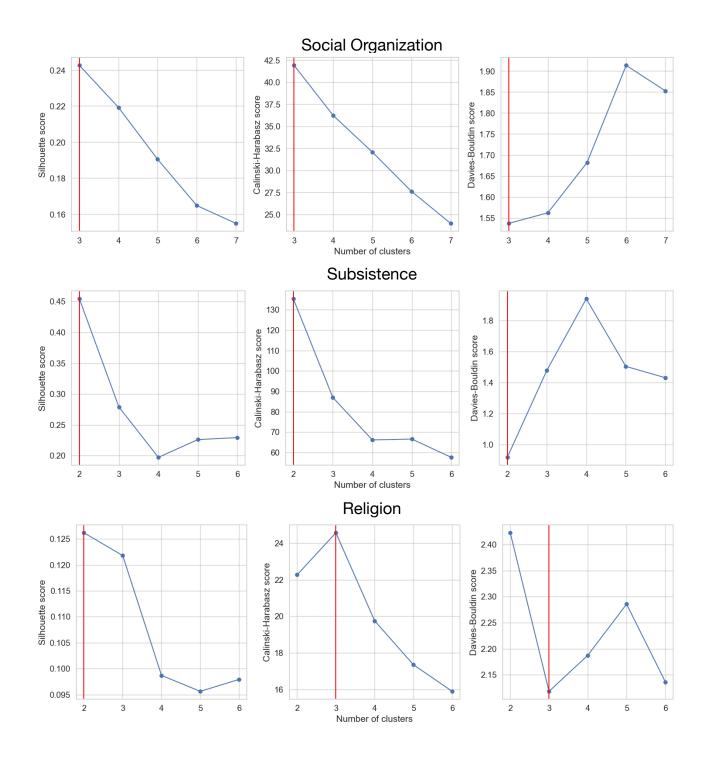


Figure S18: Kmeans clustering metrics for Hamming distances. Optimal numbers of Kmeans clusters for each data class (rows) according to the silhouette score (column 1), the variance ratio criterion (column 2), and the Davies-Bouldin score (column 3). The Hamming distance underestimates the number of clusters.

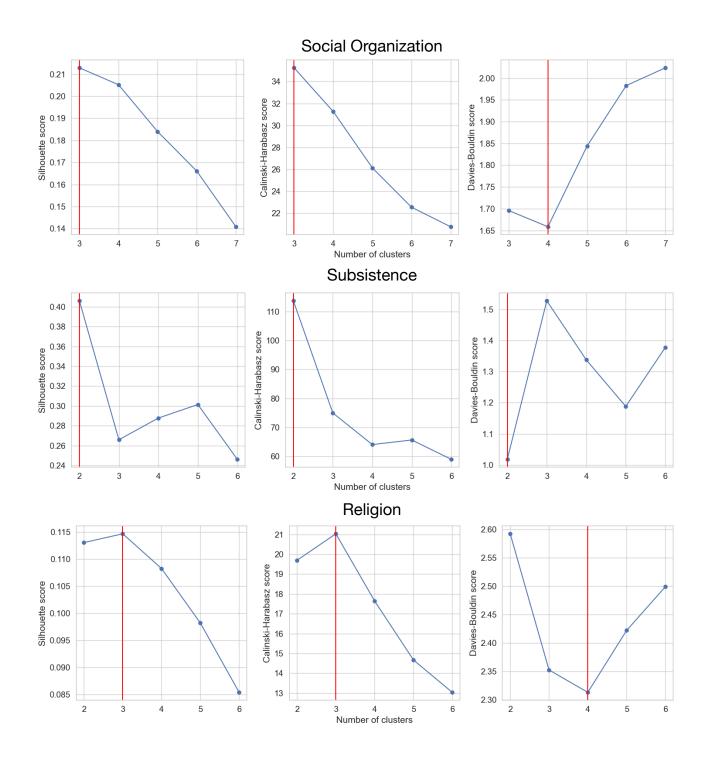


Figure S19: Kmeans clustering metrics for Jaccard distances. Optimal numbers of Kmeans clusters for each data class (rows) according to the silhouette score (column 1), the variance ratio criterion (column 2), and the Davies-Bouldin score (column 3). The Jaccard distance underestimates the number of clusters.

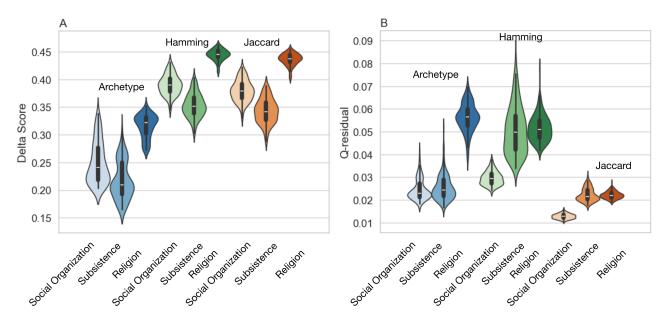


Figure S20: Comparison of δ -score and Q-residuals. The δ -score preserves relative differences in tree-likeness across choice of pairwise distance, but the Q-residual does not.

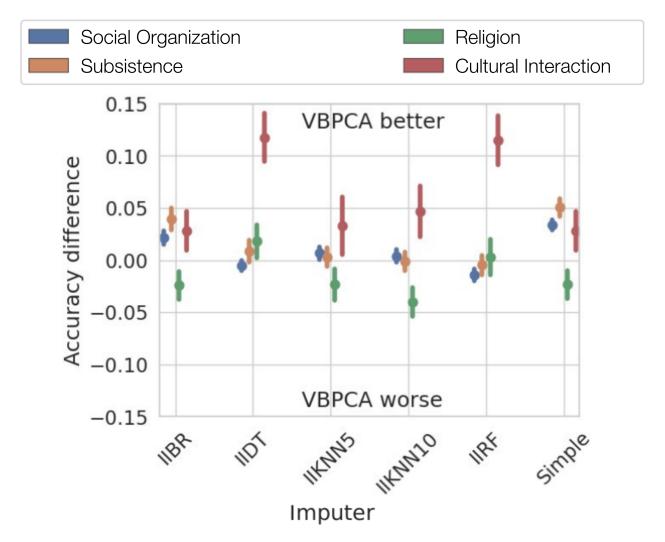
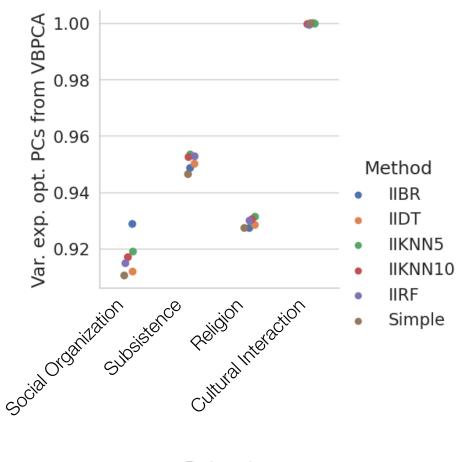


Figure S21: Imputation performance comparison. Accuracy of VBPCA imputation relative to other imputation algorithms across 50 replicates, each with 2% values deleted randomly from one of the four cultural class datasets. Given that none of the imputers consistently outperforms VBPCA and that VBPCA provides marginal posterior estimates for the data matrix (which other imputers do not), VBPCA is used for the remainder of our analysis. Positive values indicate that VBPCA performs better. The vertical lines are 95% confidence intervals for the difference in accuracy. The alternative imputers are Iterative imputer with one of the following regressors: Bayesian ridge regressor (IIBR), Iterative imputer with decision tree regressor (IIDT), Iterative imputer with nearest neighbor regressor with k = 5 (IIKNN5) and k = 10 (IIKNN10), Iterative imputer random forest regressor (IIRF), and simple imputation with missing data replaced by column means (Simple).



Dataset

Figure S22: VBPCA noise filtering for different datasets. The proportion of cumulative variance explained by the optimal number of principal components for data reconstruction as determined by VBPCA using classical PCA after filling missing values via different imputation approaches.

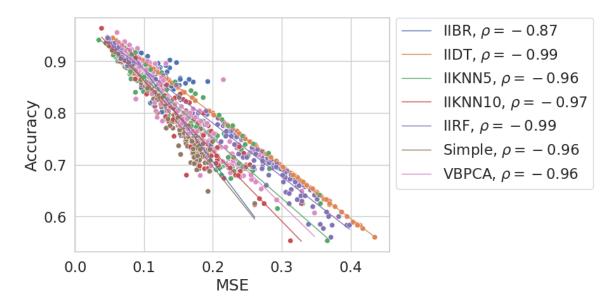


Figure S23: The relationship between Accuracy and MSE. Across all datasets and imputers, there is a strong correlation between MSE (mean squared error) and the accuracy of the inferred values.

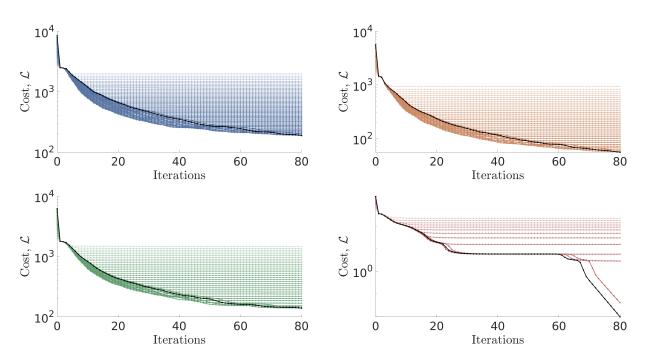


Figure S24: Cost functions for each of the four cultural classes. The colors of the lines get darker as the number of principal components increases, and the optimal number of PCs (49, 37, 34, 13 for each of the datasets) is colored in black. As the optimum is approached, the values of the cost function for a given number of principal components begin to 'stack up.'

Region	Sample mean	Standard error
Malesia	5.71	2.71
Polynesia	747.70	220.74
Papuasia	9.88	3.41
Outer Melanesia	56.50	15.69
Taiwan	7.75	7.75
Micronesia	236.51	46.12
Madagascar	0.00	0.00
Mainland SE Asia	0.00	_

Table S1: Distance to nearest neighbor (km) by geographic region. A value of 0 indicates immediate adjacency to the nearest neighbor.

Table S2: Likelihood ratio test for regional archetype distributions for geographic regions containing Polynesian outliers and Polynesia.

Archetype set	Region	LR statistic	p-value
Social Organization	Papuasia	51.0	9×10^{-10}
	Micronesia	36.4	8×10^{-07}
	Outer Melanesia	45.1	1×10^{-8}
Subsistence	Papuasia	32.3	2×10^{-6}
	Micronesia	11.9	0.0183
	Outer Melanesia	22.9	0.00013
Religion	Papuasia	38.5	9×10^{-8}
	Micronesia	17.1	0.0018
	Outer Melanesia	14.7	0.0055
Cultural Interaction	Papuasia	39.4	2.7×10^{-9}
	Micronesia	1.3	0.53
	Outer Melanesia	2.9	0.23

Table S3: Pairwise comparisons of F_{st}/F_{st}^{max} for Social Organization archetypes using Wilcoxon rank sum test with continuity correction.

	Papuasia	Micronesia	Outer Melanesia
Micronesia	< 2e-16	-	-
Outer Melanesia	< 2e-16	< 2e-16	-
Polynesia	< 2e-16	< 2e-16	< 2e-16

Table S4: Pairwise comparisons of F_{st}/F_{st}^{max} for *Subsistence* archetypes using Wilcoxon rank sum test with continuity correction.

	Papuasia	Micronesia	Outer Melanesia
Micronesia	< 2e-16	-	-
Outer Melanesia	< 2e-16	< 2e-16	-
Polynesia	< 2e-16	< 2e-16	< 2e-16

Table S5: Pairwise comparisons of F_{st}/F_{st}^{max} for *Religion* archetypes using Wilcoxon rank sum test with continuity correction.

	Papuasia	Micronesia	Outer Melanesia
Micronesia	0.7	-	-
Outer Melanesia	< 2e-16	< 1.4e-08	-
Polynesia	< 2e-16	< 2e-16	< 2e-16

Table S6: Pairwise comparisons of F_{st}/F_{st}^{max} for *Cultural Interaction* archetypes using Wilcoxon rank sum test with continuity correction.

	Papuasia	Micronesia	Outer Melanesia
Micronesia	0.0028	-	-
Outer Melanesia	< 2e-16	< 2e-16	-
Polynesia	< 2e-16	1.3e-14	3.3e-11

Table S7: Normalized F_{st} values for regions containing Polynesian outliers and Polynesia itself.

Archetype set	Region	F_{st}/F_{st}^{max}	sample size	number archetypes
Social Organization	Polynesia	0.268	18	5
	Micronesia	0.571	19	5
	Outer Melanesia	0.478	15	5
	Papuasia	0.360	29	5
Subsistence	Polynesia	0.338	18	4
	Micronesia	0.253	19	4
	Outer Melanesia	0.456	15	4
	Papuasia	0.565	29	4
Religion	Polynesia	0.560	18	4
	Micronesia	0.352	12	4
	Outer Melanesia	0.392	20	4
	Papuasia	0.339	31	4
Cultural Interaction	Polynesia	0.281	18	2
	Micronesia	0.366	12	2
	Outer Melanesia	0.194	20	2
	Papuasia	0.401	31	2

Table S8: Mann-Whitney U test, the difference in log-likelihood; alternative hypothesis Polynesian likelihood is greater. * indicates statistical significance.

Archetype set	Culture	MW statistic	p-value
Social Organization	Ellice* (Micronesia)	28508	9×10^{-14}
	Kapingamarangi* (Micronesia)	24235	0.00012
	Ontong Java* (Papuasia)	26453	1×10^{-8}
	Rennell Islanders* (Papuasia)	36452	3×10^{-46}
	Rotumans* (Outer Melanesia)	27227	2×10^{-10}
	Tikopia (Outer Melanesia)	16235	0.999
Subsistence	Ellice	18350	0.923
	Kapingamarangi	9924	1.0
	Ontong Java*	32412	3×10^{-27}
	Rennell Islanders*	27560	3×10^{-11}
	Rotumans*	21982	0.0433
	Tikopia*	26947	9×10^{-14}

Archetype set	Culture	MW statistic	p-value
Religion	Anuta (Outer Melanesia)	18451	0.910
	Bellona (Papuasia)	13909	1.0
	Aniwa (Outer Melanesia)	15734	1.0
	Kapingamarangi	10189	1.0
	Nukuoro (Micronesia)	14961	1.0
	Ontong Java	12671	1.0
	Rennell	21614	0.0815
	Rotuma	18600	0.887
	Tikopia	21826	0.0572
Cultural Interaction	Bellona	713	1.0
	Ontong Java*	36648	3×10^{-47}
	Rennell	1071	1.0

Table S9: Mann-Whitney U test, the difference in log-likelihood; alternative hypothesis Polynesian likelihood is greater. * indicates statistical significance.

Table S10: Differences in relative verticality of transmission among cultural archetype sets as indicated by Dunn's test for pairwise differences with Holm's correction for the p-values.

Metric		Social Organization	Subsistence	Religion
	Social Organization	1.0	$p < 1 \times 10^{-5}$	$p < 1 \times 10^{-5}$
δ -score	Subsistence	$p < 1 \times 10^{-5}$	1.0	$p < 1 \times 10^{-5}$
	Religion	$p < 1 \times 10^{-5}$	$p < 1 \times 10^{-5}$	1.0
	Social Organization	1.0	0.372	$p < 1 \times 10^{-5}$
Q-residual	Subsistence	0.372	1.0	$p < 1 \times 10^{-5}$
	Religion	$p < 1 \times 10^{-5}$	$p < 1 \times 10^{-5}$	1.0

Table S11: Comparison of proportion of missing data within different archetypes and for linguistic outliers. Alone among archetypes, cultures associated with the Agriculture archetype are missing large numbers of features.

Category	Median (in cat.)	Median (out of cat.)	P-value
Social Organization			
Outlier	0.083	0.071	0.68
Ambilineal Chiefdoms	0.095	0.060	0.48
Duolateral Communities	0.071	0.095	0.26
Matrilineal Clans	0.083	0.071	0.85
Neo-local Kindreds	0.024	0.083	0.37
Patrilineal Clans	0.048	0.071	0.19
Subsistence			
Outlier	0.074	0.074	0.87
Agriculture	0.78	0.074	0.00
Fish & Fruit	0.00	0.074	0.026
Grain & Cattle	0.074	0.074	0.32
Roots & Pigs	0.074	0.074	0.18
Religion			
Outlier	0.065	0.065	0.96
Animisim	0.097	0.065	0.4
Autonomous Communities	0.097	0.065	0.25
Heroes	0.065	0.065	0.83
Mana	0.032	0.097	0.022
Cultural Interaction			
Outlier	0.00	0.00	0.19
Contact	0.00	0.00	0.46
Isolation	0.00	0.00	0.46

Table S12: Correlation of missingness with extreme archetype values. In general the degree of missingness does not have a strong effect on the max/min for a given society and cultural class.

	ρ (Max)	P-value (Max)	ρ (Min)	P-value (Min)
Social Organization	0.08	0.37	-0.01	0.91
Subsistence	0.10	0.24	0.075	0.394
Religion	-0.21	0.013	0.14	0.10
Cultural Interaction	0.037	0.67	-0.037	0.67

Variable	Definition	Pagel's λ	Log Likelihood	P-value
Pul8	Primordial pair	1.00	-38.6	1E-13
Pul36	Headhunting	0.517	-62.0	4E-05
Pul21	Mana as a spiritual or religious concept	0.497	-60.1	1E-07
Pul84	Religious Authority	0.394	-148.9	0.003
Pul18	Mana related to social influence or technical skill	0.392	-60.4	2E-05
Pul2	Belief in god(s)	0.389	-133.6	0.0005
Pul3	Belief in nature god(s)	0.379	-130.6	0.009
Pul10	The actions of others after one has died	0.339	-93.6	0.0002
	can affect the nature of one's afterlife			
Pul35	Costly sacrifices and offerings	0.330	-29.3	0.0115
Pul4	Belief in deified ancestor(s)	0.309	-148.6	0.0037
Pul37	Political and religious differentiation (SCCS v 757)	0.298	-109.9	0.0118
Pul20	Mana as a personal quality	0.288	-67.4	0.0007
Pul41	Tattooing	0.288	-121.3	0.0388
Pul38	Largest religious community	0.281	-126.5	0.004
Pul85	Largest religious community	0.250	-152.4	0.0240
Pul5	Belief in ancestral spirits	0.214	-115.6	0.107
Pul39	Genital cutting	0.209	-140.4	0.3
Pul11	Own actions affect own afterlife	0.164	-90.6	0.5
Pul14	Social hierarchy tapu	0.151	-63.9	0.108
Pul42	Scarification	0.094	-80.7	0.5
Pul16	Resource management tapu	0.083	-72.7	0.4
Pul12	Belief in culture hero(es)	0.034	-128.6	0.7
Pul7	Belief in supernatural punishment for impiety	0.025	24.0	0.7
Pul6	Belief in nature spirits	7E-05	-121.8	1
Pul9	Myth of humanity creation	7E-05	-119.9	1
Pul15	Kinship tapu	7E-05	-20.5	1
Pul40	Tooth pulling	7E-05	-117.5	1
Pul43	Piercing	7E-05	-97.8	1

Table S13: *Religion* traits sorted by phylogenetic signal over the linguistic phylogeny from Gray et al. (2009). P-values are from LRT with null assumption $\lambda = 0$.

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