

1 **Hunter-gatherers adjust mobility to maintain contact under climatic variation**

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3 **Matt Grove**

4 Archaeology, Classics and Egyptology

5 University of Liverpool

6 12-14 Abercromby Square

7 Liverpool L69 7WZ

8 matt.grove@liverpool.ac.uk

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33 **Abstract**

34 Population density and mobility are fundamental population parameters for hunter-gatherer groups,
35 and their reconstruction for prehistoric populations has long been an aim of archaeological research.
36 This endeavour has become more important than ever in recent years, with the recognition that
37 these parameters play a key role in determining rates of cultural transmission. Potential
38 archaeological proxies for population density and mobility are often hard to interpret, creating a
39 need for more generic, reliable, and easily calculated indicators. Climatic variables provide
40 considerable promise in this area, and the analyses reported here test the efficacy of six climatic
41 variables as potential predictors. Significant predictors are then incorporated in path analyses that
42 assess the causal relationships between climatic variables, population density, and mobility. Results
43 suggest that the previously established strong reciprocal relationship between population density
44 and mobility is not due purely to common determination by climatic variables. Instead, the best
45 supported model is consistent with the hypothesis that hunter-gatherers adjust levels of mobility so
46 as to maintain contact with neighbouring groups at varying population densities. This ensures that
47 opportunities for cultural transmission are maintained at similar levels regardless of climatic
48 variation. The results lead to a number of archaeologically testable predictions concerning the
49 relationships between climatic variables, population density, mobility, and assemblage complexity.

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51 **Keywords:** hunter-gatherer; climatic variability; mobility; population density; cultural transmission.

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64 **Hunter-gatherers adjust mobility to maintain contact under climatic variation**

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66 **1. Introduction**

67 As well as being fundamental population parameters themselves, population density and mobility
68 are heavily implicated in patterns of cultural transmission (e.g. Powell et al. 2009; Grove 2016). For a
69 given level of mobility, individuals or groups experiencing higher population density will encounter
70 one another more frequently per unit time. Similarly, at a given population density, more mobile
71 individuals or groups will encounter one another more frequently per unit time. Thus increases in
72 the number of individuals per unit area or the area covered per unit time will increase the encounter
73 rate, with each encounter providing an opportunity for the transmission of information. The
74 estimation of population densities and levels of mobility associated with archaeological sites is
75 therefore of considerable importance to both studies of palaeoecology and research into rates of
76 cultural transmission.

77 Variation in population density and mobility is generally considered to reflect underlying differences
78 in climatic variables that account for the density of subsistence resources, and most studies of
79 hunter-gatherers support such conclusions (e.g. Kelly 1995; Binford 2001; Grove 2009). High
80 resource densities should sustain high population densities and require minimal mobility, whilst low
81 resource densities should sustain only low population densities, requiring greater mobility. In
82 addition to these basic expectations, mobility is viewed as a means of maintaining equitable
83 relations with distant groups to ensure that they can be called upon to share resources in times of
84 local hardship. The existence of such 'safety nets' is a common theme in considerations of hunter-
85 gatherer mobility strategies (e.g. Wiessner 1982; Whallon 2006), and reinforces the link between
86 demographic variables and underlying resource structure.

87 A growing number of archaeologists are making use of demographic variables in explaining changes
88 in assemblage diversity, complexity, or sophistication through time or across space (e.g. Brumm and
89 Moore 2005; James and Petraglia 2005; Zilhao 2007; Langley et al. 2011). Much of this work is
90 influenced by the theoretical models of Shennan (2001) and Henrich (2004), which relate population
91 size to the 'cultural fitness' or 'skill level' attainable within that population. Henrich's (2004) analysis
92 features a particularly effective model of the cultural transmission process, and has had considerable
93 impact within archaeology. The model introduces a basic sampling problem in which all members of
94 a population copy a target individual with a degree of error. Most copying errors are detrimental,
95 but a small number will be beneficial; a larger population size therefore increases the probability
96 that at least one member of the population will produce a copy that is better than the original, thus
97 ensuring cumulative improvement in technology.

98 Powell and colleagues (2009) embedded the Henrich (2004) model within a broader simulation
99 framework, demonstrating that population density and mobility are more realistic correlates of
100 cultural sophistication than is population size. Grove (2016) confirmed the lack of a population size
101 effect, and noted that both population density and mobility are proxies for the rate of encounters
102 between individuals. The encounter rate model both unifies previous models and explains why
103 empirical studies that focus on population density alone are unlikely to find significant effects. In
104 parallel, theoretical modelling by Premo (2012) has emphasized the 'connectedness' of populations

105 as playing a key role in the resilience of their material culture traditions. An intriguing archaeological
106 case study in this vein is provided by Riede (2008, 2016), who argues that the eruption of the
107 Laacher See volcano in western Germany around 12,920 BP had considerable implications for Late
108 Glacial human demography in northern Europe. In particular, Riede (2008:596) postulates that a
109 “significant and sudden” reduction in the connectedness of populations led to the loss of bow-and-
110 arrow technology and regional technological simplification. An important strand linking both
111 theoretical and empirical work is that decreases in population density, mobility, or ‘connectedness’
112 can lead to decreases in the pool of cultural interactants, thus limiting the potential for cumulative
113 cultural evolution.

114 Despite the growing use of demographic arguments, few studies link demographic variables to
115 climatic variables, despite the relatively well established patterns emerging from the hunter-
116 gatherer literature. This omission is part of a wider problem in that few studies examine proxies for
117 demographic variables in parallel with evidence for variation in assemblage sophistication.
118 Hypotheses linking demography with cultural transmission are therefore frequently employed as
119 explanations, but rarely tested with archaeological data (see Collard et al. 2016). One possible
120 reason for this trend is that, whilst there are numerous aspects of archaeological assemblages that
121 might act as proxies for demographic variables, they are often subject to confounds that are difficult
122 to eliminate. Raw material or artefact transfer distances might act as proxies for mobility (e.g.
123 Taborin 1993; Féblot-Augustins 1997; Pearce and Moutsiou 2014), but in many cases it is impossible
124 to distinguish between the signatures of mobility and trade. Artefact or faunal densities, particularly
125 in landscape-level studies, may correlate with population densities, but will also correlate with
126 localised group sizes and site occupation durations (e.g. Grove 2009; Tryon and Faith 2016). The
127 frequent occurrence of palimpsest data and the difficulties of establishing true contemporaneity
128 between sites serve to further cloud this picture (Schacht 1984; Stern 1993; Grove 2011).

129 The study of relationships between basic climatic and demographic variables in hunter-gatherer
130 groups has the potential to considerably improve our understanding of prehistoric demography. Not
131 only can such relationships provide baseline information about likely demographic parameters, but
132 they can also be used to verify inferences drawn directly from archaeological assemblages, providing
133 independent and complimentary lines of evidence (*sensu* Wylie 1989). Given the ever wider
134 availability of high resolution palaeoclimatic data, it is becoming increasingly important to fully
135 explore the structure of relationships between climatic variables, population density and mobility in
136 recent hunter-gatherer groups. This will not only facilitate inferences about prehistoric
137 demographics, but will also enable archaeologists to make more informed, robust conclusions about
138 the potential for cultural transmission in prehistoric societies.

139 Given the theoretical relationships between resource density, population density, and mobility
140 highlighted above in relation to the potential for cultural transmission, it is important to distinguish
141 between three viable hypotheses concerning the interaction of these variables. Grove (2016)
142 empirically demonstrated a strong negative correlation between population density and mobility in
143 hunter-gatherer populations. This result was interpreted in a social context as suggesting that
144 mobility is increased as population density decreases, to ensure maintenance of a sufficient number
145 of encounters between neighbouring groups. However, the causal links between resource density,
146 population density and mobility were not directly tested in Grove (2016). There remain two further
147 hypotheses that could feasibly account for the observed patterning. Firstly, the relationship between

148 population density and mobility could result purely from their common determination by resource
149 density. When resources are at low density, foragers dependent on those resources will have their
150 population densities constrained; similarly, they will be required to cover a greater area in order to
151 fulfil energetic requirements. Secondly, if groups are required to cover greater areas to obtain
152 resources, they may increase distances between groups to avoid depleting the same areas as their
153 neighbours, thus decreasing population density.

154 The analyses detailed below therefore test three alternative hypotheses:

155 **H1.** The relationship between population density and mobility is due purely to their common
156 determination by resource density.

157 The two other hypotheses can be most succinctly framed as different responses to decreasing
158 resource density:

159 **H2.** Groups increase mobility to maintain contact with neighbouring groups who are forced by low
160 resource densities to live further apart (i.e. at lower population densities);

161 **H3.** Groups decrease population density (i.e. move further apart) to avoid depleting the same areas
162 as neighbouring groups who are forced by low resource densities to increase mobility.

163 These hypotheses differ only with respect to the *direct* effects of resource density. H1 suggests that
164 resource density directly affects both population density and mobility; H2 suggests that resource
165 density affects population density directly but mobility only indirectly; H3 suggests that resource
166 density affects mobility directly but population density only indirectly. A schematic of these direct
167 and indirect causal paths is provided in Figure 1.

168 It is important to distinguish between these three hypotheses as they have a direct bearing on how
169 climatic variables are likely to affect rates of cultural transmission in prehistory. Specifically, the
170 structure of dependence between these variables affects the extent to which opportunities for
171 cultural transmission are seen as imposed directly by climatic variables as opposed to arising from
172 social responses to those variables. Whilst feedback mechanisms between climatic and social
173 systems are inevitable, and the former can be seen as ultimately constraining the possibilities of
174 habitation, it is often argued that the latter specify the ways in which climatic constraints are
175 addressed by a particular society (e.g. Ingold 1981; Gamble 1986; Kelly 1995). In these terms,
176 support for H1 would suggest that social systems – in as far as they are defined by basic variables
177 such as population density and mobility – are heavily constrained by climatic variables. Support for
178 H2 or H3, by contrast, would suggest a greater measure of flexibility in how human societies meet
179 climatic challenges.

180 H1 can be assessed purely via partial correlation analyses; if the strong relationship between
181 population density and mobility remains when the effects of resource density on both are controlled
182 for, H1 is not supported. The causal structures of H2 and H3, however, immediately suggest path
183 analysis as the appropriate statistical technique. Path analysis is rarely used in archaeology and
184 anthropology, but is ideal when evaluating hypotheses that involve multiple causal steps (see Wright
185 1921, 1934; Li 1975). It is a relatively simple extension of regression analysis, and can be paired with
186 model selection statistics to allow powerful discrimination between competing hypotheses.

187 **2. Data**

188 There exists no direct measure of global variation in resource density, but a number of climatic
189 variables can be calculated that act as useful indices. In particular, a long history of research into
190 biome classification has established that the productivity of different biomes is determined primarily
191 by levels of temperature and precipitation (e.g. Holdridge 1947; Whittaker 1975; Olson et al. 2001;
192 Walter and Breckle 2002; Cox et al. 2016). In order to adequately represent resource density, data
193 on four primary climatic variables (mean annual temperature, standard deviation of annual
194 temperature, mean annual precipitation and standard deviation of annual precipitation) were
195 collected. Two composite variables were also calculated. Net above-ground productivity (NAGP)
196 provides an estimate of net primary productivity (the mass of new plant growth per year) from a
197 combination of temperature and precipitation measures (Rosenzweig 1968). Effective temperature
198 (ET) employs a global model to simultaneously represent both the length and the warmth of the
199 growing season (Bailey 1960). These six climatic variables were combined with data on hunter-
200 gatherer population density and mobility.

201 Data on population density, mobility, and net above-ground productivity (NAGP) for the hunter-
202 gatherer sample were gathered from Binford (2001). Binford's (2001) full sample was reduced to
203 only those groups that are fully mobile (i.e. those for which his GRPPAT variable is equal to 1),
204 yielding a sample of 175 groups. Mobility (distance moved per annum) was converted from miles to
205 kilometres prior to analysis, and the population density variable was converted from individuals per
206 100km² to individuals per km².

207 Climatic variables, with the exception of NAGP, were calculated from the global high-resolution
208 databases archived at NOAA/OAR/ESRL PSD (Boulder, Colorado, USA, see
209 <http://www.esrl.noaa.gov/psd/> and Willmott and Matsuura 2001). These databases provide monthly
210 mean measurements of air temperature (interpolated from 12,587 stations) and precipitation
211 (interpolated from 12,857 stations) on a .5 × .5 degree global grid. Mean annual temperature (MAT),
212 mean annual precipitation (MAP), standard deviation in annual temperature (SDT) and standard
213 deviation in annual precipitation (SDP) were calculated via linear interpolation of the 12 monthly
214 mean temperatures in the global grid using the longitude and latitude data given in Binford (2001).
215 Finally, effective temperature (ET) was calculated following Bailey (1960:4) as $ET = (18\bar{w} -$
216 $10\bar{c}) / (8 + \bar{w} - \bar{c})$, where \bar{w} denotes the mean temperature of the warmest and \bar{c} the mean
217 temperature of the coldest month of the year.

218 To satisfy the requirement of the statistical techniques employed below for normally distributed
219 data, variables with substantial positive skew (population density, MAP, SDP, ET, and NAGP) were
220 natural log transformed, and variables with moderate positive skew (mobility and SDT) were square
221 root transformed. MAT showed moderate negative skew and was transformed using the equation
222 $t_i = -\sqrt{u_{\max} - u_i}$, where t is the transformed variable and u the raw variable (Tabachnick and
223 Fidell 2007). All variables were then standardized (i.e. converted to z-scores) prior to analysis, as this
224 prevents the need to calculate intercepts, increasing the accuracy of the path analyses (Li 1975) and
225 the reliability of model comparisons. It further entails that the standardized and unstandardized
226 regression coefficients are identical (i.e. $B = \beta$).

227

228 **3. Methods**

229 The first stage of the analysis involved performing two separate multiple regressions of (1)
230 population density and (2) mobility on the six climatic variables. Multiple regressions were run using
231 the 'enter' routine in IBM SPSS Statistics 24; this routine produces partial regression coefficients for
232 each independent variable, making the results identical to those achieved via simple path analysis.
233 Confidence intervals for all regression (β) and correlation (r) coefficients were produced using SPSS.
234 Due to the functionality of the software confidence intervals for regression coefficients were
235 produced analytically, whereas for bivariate and partial correlation coefficients they were produced
236 via bootstrap sampling. All bootstrap confidence intervals were produced from 10,000 samples.

237 To assess H1, climatic variables found to be significant predictors of either mobility or population
238 density were included as control factors in partial correlations. A result demonstrating that the
239 correlation between population density and mobility becomes non-significant when the relevant
240 climatic variables are controlled for is consistent with H1. Conversely, if the partial correlation
241 remains significant, H1 is rejected. Comparing the simple bivariate correlation between the two
242 variables with the partial correlation gives an impression of the proportion of the correlation
243 accounted for by the climatic variables.

244 To assess H2 and H3, variables found to be significant predictors of either mobility or population
245 density were included in path analyses that were set up as two sets of causal models. The models in
246 Set 1 represent H2, that climatic variables affect mobility primarily indirectly, via their effects on
247 population density. The models in Set 2 represent H3, that climatic variables affect population
248 density primarily indirectly, via their effects on mobility. Differences between models in a given set
249 are due to different configurations of paths from the exogenous (climatic) to the endogenous
250 (population density and mobility) variables. Regardless of the configuration of the climatic variables,
251 models comprising Set 1 include a directed path from population density to mobility, whereas
252 models comprising set 2 include a directed path from mobility to population density. It is expected
253 that increases MAP, MAT, NAGP and ET will lead to increases in resource density, and therefore to
254 increases in population density and reductions in mobility. The variables that index variation in
255 temperature and precipitation – SDT and SDP respectively – are more complex, but in line with
256 previous research (Dyson-Hudson and Smith 1978; Winterhalder 1986; Kelly 1995; Johnson and Earle
257 2000) greater climatic variation is expected to reduce population density and increase mobility.

258 For each model a log-likelihood, a sample size corrected AICc value, and a delta value were
259 calculated; these in turn enabled the calculation of relative model likelihoods and probabilities given
260 the data and the set of models considered (see Burnham & Anderson 2002; Burnham et al. 2011).
261 This method allows not only for the identification of the most supported model, but also for
262 assessment of the relative merits of all the models comprising Set 1 as opposed to all the models
263 comprising Set 2. Path analyses were conducted in IBM SPSS Amos 24, with additional calculations
264 based on the raw AIC values and parameter numbers output by the program, following Burnham
265 and Anderson (2002).

266 **4. Results**

267 A multiple regression of population density on the six climatic variables found that only SDT ($\beta = -$
268 $.734$ [95% CI: $-1.013, -.454$], $p < .001$) and SDP ($\beta = .433$ [$.190, .675$], $p < .001$) were significant

269 predictors. Isolating these two variables demonstrated that together they explain 49.4% of the
270 variance in population density ($R^2 = .494$, $F(2,173) = 84.379$, $p < .001$). The multiple regression of
271 mobility on these six variables found that, again, only SDT ($\beta = .885$ [.582, 1.188], $p < .001$) and SDP
272 ($\beta = -.435$ [-.698, -.173], $p = .001$) were significant predictors. Isolating these two variables
273 demonstrated that together they explain 38.0% of the variance in mobility ($R^2 = .380$, $F(2,173) =$
274 53.059 , $p < .001$).

275 The assessment of **H1** involved first calculating the simple bivariate correlation between population
276 density and mobility, yielding $r = -.830$ [-.913, -.747], $n = 175$, $p < 0.001$. As only SDT and SDP were
277 found to be significant predictors of population density and mobility, partial correlations were run
278 controlling for SDT, controlling for SDP, and controlling for both. As a final verification, a partial
279 correlation was run controlling for all six climatic variables. Bootstrap confidence intervals for r
280 (10,000 samples) were produced for each partial correlation. The results, presented in Table 1,
281 demonstrate that the correlation between population density and mobility remains strong and
282 significant in all cases, rejecting the hypothesis that this correlation is due purely to common
283 determination by the climatic variables.

284 The finding that SDT and SDP were the only significant predictors of both PD and DISMOV made it
285 possible to run an exhaustive series of exploratory path analyses on the two exogenous variables
286 (SDT and SDP) and two endogenous variables (population density and mobility). The twelve possible
287 models are shown in Figure 2; Models 1-6 comprise Set 1 and represent variations on H2, whereas
288 Models 7-12 comprise Set 2 and represent variations on H3. Model 2 represents the strongest
289 embodiment of the hypothesis that the climatic variables affect mobility only indirectly (H2);
290 similarly, model 8 represents the strongest embodiment of the hypothesis that climatic variables
291 affect population density only indirectly (H3).

292 Results of the path analyses on the 12 models are shown in Table 2, with full path diagrams for the
293 four best models presented in Figure 3. The first result to note is that models 1 and 7 are saturated
294 (i.e. the maximum possible number of paths are present), and thus are not amenable to reliable
295 statistical analysis. For the remaining models, it is important to note that better models have low
296 AICc and delta values and high relative likelihoods (L_i) and probabilities (w_i). Thus model 2 is the
297 best model, and model 8 is the worst model. The evidence ratio, calculated by dividing the
298 probability of model 2 by that of model 8 (see Burnham et al. 2011), demonstrates that the empirical
299 support for model 2 is over 44 million times that for model 8.

300 Assessing the two sets of models provides more conservative results. Summing the probabilities of
301 the five valid Set 1 models (2-6) and the five valid Set 2 models (8-12) and dividing the former by the
302 latter demonstrates that the empirical support for Set 1 is ≈ 5.86 times that for Set 2. This relatively
303 moderate evidence ratio is due to the performance of model 11, which is the only reasonably
304 supported model from Set 2. However, comparing model 11 to its topological equivalent from Set 1
305 (model 3) demonstrates that the latter receives ≈ 2.68 times more support. In summary, the results
306 support Set 1 over Set 2, and therefore support the hypothesis that climatic variables affect mobility
307 primarily indirectly, via their effects on population density (H2). Support is strongest for model 2,
308 which is the simplest and strongest embodiment of this hypothesis.

309

310 5. Discussion

311 The results presented above provide the most support for the hypothesis that climatic variables
312 determine population density, which in turn determines mobility (H2). However, they also suggest
313 that it is primarily changes in the standard deviations of temperature and precipitation rather than
314 changes in their means that affect these demographic variables. That variables indexing climatic
315 variability provide the key to demographic patterns is an important conclusion, suggesting not only
316 that such variability has a powerful effect on sustainable population densities, but also that contact
317 between neighbouring groups is crucial in mitigating the challenges it creates. Surprisingly, ET and
318 NAGP, which are often considered to be good indicators of resource density (e.g. Kelly 1995; Binford
319 2001; Grove 2009), play no part in the final models. Whilst the relationship between SDT and
320 population density is as predicted, the relationship between SDP and population density runs
321 counter to the prediction, with increasing variability in precipitation found to *increase* population
322 density. This result is surprising, and merits further scrutiny.

323 One possible explanation arises from the fact that the relationship between SDT and SDP is itself
324 significant and negative ($r = -.734 [-.787, -.675]$, $df = 173$, $p < .001$). At the locations represented by
325 the hunter-gatherer groups within the sample, higher variability in temperature tends to be
326 associated with lower variability in precipitation. It could be argued, therefore, that higher SDP leads
327 to higher PD simply because both show significant positive relationships with SDT. Note, however,
328 that this goes against the structure of the models presented above; if all the explained variability in
329 PD were due purely to variation in SDT, then SDP would not achieve significance as a predictor. To
330 reinforce this point, a partial correlation of SDP and PD was performed whilst controlling for SDT.
331 The resultant correlation remains significant and positive ($r = .236 [.086, .379]$, $df = 172$, $p = .002$),
332 suggesting that this explanation is not sufficient.

333 A second possible explanation arises from the positive correlation between SDP and MAP ($r = .859$
334 $[.783, .936]$, $df = 173$, $p < .001$), suggesting that perhaps high SDP leads to high population densities
335 simply because it correlates with MAP, higher values of which were expected to lead to higher
336 population densities. To test this possibility, a partial correlation of SDP and PD was performed
337 whilst controlling for the effects of MAP on both variables. The resultant correlation remains
338 significant and positive ($r = .321 [.199, .429]$, $df = 172$, $p < .001$), suggesting that the positive
339 relationship between MAP and SDP alone is not sufficient to explain the finding of higher variability
340 in SDP leading to higher population densities.

341 Finally, it may be that the widely assumed link between lower SDP and higher resource density is
342 itself inaccurate or mistaken. Some authors have argued that resource *diversity* might be more
343 important to hunter-gatherer groups than resource density (e.g. Yesner 1977; Erlandson 1994), and
344 it is not unreasonable to suspect that greater variability in precipitation might lead to a greater
345 diversity of floral resources, and hence a wider range of secondary consumers. Further to this, SDP is
346 a valuable index of, but does not fully reflect, differences in seasonality between regions at the
347 macro-scale. The four seasons of the temperate zones, for example, create different rainfall regimes
348 than those experienced in the tropics, which experience only two seasons, often with two distinct
349 periods of increased precipitation. Locations at widely divergent latitudes, therefore, could
350 demonstrate equivalent SDP but experience different annual patterns of rainfall, leading to different
351 relationships between SDP and population density.

352 To test this latter possibility, the data were divided into three subsets by latitude: the Tropical subset
353 includes those groups between the tropics of Cancer and Capricorn; the Temperate subset includes
354 those between the Arctic Circle and the Tropic of Cancer, and between the Tropic of Capricorn and
355 the Antarctic Circle; finally, the Polar subset includes those north of the Arctic Circle or south of the
356 Antarctic Circle. Population density was then regressed on SDP separately for each of these three
357 subsets. The results, shown in Table 3, demonstrate that SDP has less of an effect on population
358 density among Tropical hunter-gatherers than it does within the other two subsets. The
359 standardized coefficients, however, remain positive and significant in all cases, suggesting that
360 greater variability in precipitation does indeed lead to greater population densities at all latitudes. In
361 future studies, however, it may prove useful to further quantify the profiles of seasonal variability
362 that exist at different latitudes, as these may reveal finer-grained patterning.

363 The results shown in Table 2 demonstrate that model 2 is the best model, and together with the
364 path diagrams of Figure 3 allow some useful comparisons to be made with the second, third, and
365 fourth best models. Firstly, note that models 3 and 4 are each equivalent to model 2 with an extra
366 path added (SDT → M in model 3, SDP → M in model 4). The path coefficients show that the added
367 path in model 3 explains only 0.0064% of the variance in mobility (i.e. 0.08^2), whilst the added path
368 in model 4 contributes essentially nothing to the explanation of this variance. This pattern is
369 reinforced by the results in Table 2. Since the AIC is calculated as $AIC = -2 \ln(L) + 2k$, where $\ln(L)$
370 is the log-likelihood and k is the number of parameters, adding an extra parameter to a model must
371 increase the log-likelihood by >1 to produce a better model (an equivalent but more complex logic
372 applies to the sample-size corrected AICc). Model 3 increases the log-likelihood by <1 , whilst model
373 4 does not increase it at all. Model 11 is identical to model 3 but for the inversion of the path
374 between population density and mobility; this inversion decreases the log-likelihood by 0.985 (thus
375 increasing the AIC by 1.97), demonstrating the greater explanatory power of models in which the
376 causal path leads from population density to mobility.

377 The principal finding of these analyses, that climatic variables exert direct effects on population
378 density but only indirect effects on mobility, is consistent with the hypothesis that hunter-gatherers
379 adjust mobility so as to maintain contact with neighbouring groups. This suggests that continued
380 social contact between groups over often considerable distances is a fundamental aspect of hunter-
381 gatherer adaptation; networks of contact between groups are imposed upon, not by, the
382 environmental substrate. Flexibility in mobility ensures that encounter rates between groups are
383 maintained at similar levels *regardless of climatic variation*, a fact that has important implications for
384 patterns of cultural transmission both within extant hunting and gathering groups and in
385 archaeologically documented populations.

386 A number of recent studies have focused on the elements of hunter-gatherer sociality that might
387 facilitate the spread of cultural variants (e.g. Apicella et al. 2012; Hill et al. 2014; Derex and Boyd
388 2016; Salali et al. 2016; Migliano et al. 2017). The conclusions of many of these studies mirror
389 aspects of research by Granovetter (1973; see also Watts and Strogatz 1998) that stressed the
390 importance of occasional far-reaching links to other individuals or groups. The idea of an optimal
391 balance of within-group to between-group transmission is therefore once again coming to the fore.
392 For example, Migliano and colleagues (2017) suggest that the efficiency of cultural transmission
393 depends on the structure of links between families. If each member of Family 1 establishes a link to
394 a member of Family 2, there will be considerable redundancy in information flow along those links,

395 and Family 1 will remain isolated from all families other than Family 2. By contrast, if each member
396 of Family 1 establishes a link to a *different* family (i.e. Member 1 links to Family 2, Member 2 to
397 Family 3, etc.), then Family 1 will be thoroughly enmeshed in the wider network, ensuring a
398 comprehensive flow of information from all quarters. In such cases, any innovations that occur in the
399 wider population should be rapidly transmitted to Family 1, who can proceed to adopt them if they
400 choose. In terms of cultural transmission, establishing individual links to multiple families is a better
401 strategy than establishing multiple links to a single family.

402 The above scenario is intuitively appealing, but an intriguing counterpoint is provided by the study of
403 Derex and Boyd (2016). Performances on the experimental computer task employed by these
404 authors suggest that partially connected groups will produce solutions that are both more complex
405 and more diverse than those of fully connected groups. These findings are best explained by noting
406 that greater diversity is inevitable when groups are not connected – alternate solutions should be
407 expected when groups cannot communicate their ideas – but that the few connections that *do* exist
408 allow the foremost achievements of each group to be periodically combined into more complex and
409 superior solutions. Derex and Boyd (2016) argue that in fully connected groups variation in solutions
410 is swamped by high levels of copying, such that only a small number of possible solutions are
411 actually explored. By contrast, partial connectivity results in a fuller exploration of the full space of
412 possible solutions. Thus there may be an optimal degree of connectivity that balances the need to
413 generate and maintain variation with the need to fully exploit the beneficial innovations that the
414 exploration of different solutions creates.

415 The degree of social contact between groups may therefore play a key part in the production and
416 transmission of innovations, ensuring that variation is maintained whilst simultaneously spreading
417 useful information between groups. In many ways, this concern directly mirrors the need to
418 maintain genetic diversity by ensuring that inbreeding is avoided, a problem that is particularly acute
419 in small populations. Large-scale mobility, and the links that it provides between distant groups, is an
420 important means of alleviating this problem. In a broad sample of Amazonian societies, Walker
421 (2014) demonstrated that rates of exogamy are in fact considerably higher in hunter-gatherer
422 groups than among horticulturalists, suggesting that the mechanisms of inbreeding avoidance in
423 hunter-gatherer society are highly developed. Incest taboos and exogamy rules are common
424 features of hunter-gatherer culture (e.g. Wobst 1975; Turner and Maryanski 2005), and a particularly
425 clear example is given in Headland’s (1987) study of the Casiguran Agta of the Philippines. The
426 exogamy rule among the Agta states that “one may not marry any person whom he already calls by
427 any kinship term” (Headland 1987:267); it thus prevents marriages between existing affines, and
428 ensures that any two family groups of Agta are linked by only a single matrimonial tie. More
429 importantly, it ensures that each family is linked to a number of others through matrimony; a rule
430 that is enforced in order to avoid incest thus has the additional benefit of creating a network
431 topology that displays the ‘partial connectivity’ quality identified by Derex and Boyd (2016).

432 Two recent studies have provided evidence that expanded social networks of the kind found in
433 contemporary hunter-gatherers may have considerable antiquity within human lineages. Sikora and
434 colleagues (2017) analysed the genomes of four individuals considered to be members of a single
435 human group from the Upper Palaeolithic site of Sunghir, Russia. None of the four were closely
436 related, leading the authors to conclude that they belonged to a society whose cultural rules
437 prevented significant endogamy. Levels of inbreeding were similar to those found among modern

438 hunter-gatherers, prompting Sikora and colleagues (2017:662) to propose that, by around 34 ka,
439 “complex family residence patterns, relatively high individual mobility, and multilevel social
440 networks were already in place”. Brooks and colleagues (2018) suggest that the formation of
441 networks of exchange and procurement over extended areas may date back even further in time, to
442 the roots of the Middle Stone Age. By around 300 ka at Olorgesailie, Kenya, hominin foragers were
443 transporting obsidian from seven separate sources at between 25 km and 50 km from the site. Raw
444 materials from these seven sources together constitute 78% of the obsidian sample, suggesting an
445 established, extensive network rather than occasional forays. Brooks and colleagues (2018) stress
446 the potential for such networks to provide responses to increasing climatic variation of the kind
447 shown above to directly influence population densities.

448 From an archaeological perspective, the finding that just two climatic variables explain
449 approximately 50% of the variance in hunter-gatherer population densities suggests that the
450 reconstruction of demographic variables via climatic correlates is likely to be a beneficial enterprise.
451 More importantly, the finding that mobility is adjusted so as to maintain contact between groups
452 under varying population density implies a number of archaeologically testable corollaries. Regions
453 in which palaeoclimatic data are of sufficient resolution to show high variability in precipitation and
454 low variability in temperature should support dense human populations, particularly at temperate
455 latitudes. Archaeological sites within these regions should demonstrate archaeological evidence of
456 both high population densities *and* low levels of mobility. To give one example, site densities should
457 be high, whilst lithic transfer distances should be low. Conversely, regions for which palaeoclimatic
458 data demonstrate low variability in precipitation and high variability in temperature should manifest
459 low archaeological site densities, with sites linked by high lithic transfer distances. Despite these
460 differences in population density and mobility, encounter rates should be similar, implying that
461 levels of potential assemblage sophistication or complexity should be essentially invariant to
462 differences in climatic regimes.

463 This final point could be interpreted as contradicting the prediction made by Torrence (1983, 1989),
464 that groups living at higher latitudes should develop more complex toolkits. Torrence’s (1983, 1989)
465 logic is that prey animals will be encountered less often at high latitudes due to their lower density,
466 and therefore that greater investment in technology is merited to minimize the risk of failing to
467 successfully harvest any animals that *are* encountered. This ‘risk hypothesis’, as it is often described,
468 is a hypothesis about a *cause* of toolkit complexity. The ‘encounter rate hypothesis’, as outlined here
469 and elsewhere (e.g. Henrich 2004; Powell et al. 2009; Grove 2016), is a hypothesis about a *constraint*
470 on toolkit complexity. Complex toolkits should appear in ethnological or archaeological contexts only
471 when both the constraint is obviated (through a sufficiently high number of encounters with other
472 groups) and the cause is present (due to a demonstrable need for more complex technology).
473 Though they are frequently treated as such, there is no clear rationale for regarding these two
474 hypotheses as mutually exclusive; indeed, future studies would benefit from reconciling them by
475 differentiating between the potential for producing complex artefacts and the actual requirement to
476 do so.

477 **6. Conclusions**

478 The results presented above suggest that basic climatic variables account for a considerable amount
479 of variation in hunter-gatherer population densities, and that the standard deviations of climatic

480 variables are of greater consequence than their means. Path analyses show the greatest support for
481 a model in which climatic variables influence mobility only indirectly, via their effects on population
482 density. This finding is consistent with the hypothesis that hunter-gatherers adjust levels of mobility
483 so as to maintain contact with neighbouring groups at varying population densities, a mechanism
484 that ensures encounter rates remain at similar levels regardless of climatic variation. In line with
485 previous research, this suggests that hunter-gatherers maintain an optimised level of connectivity
486 between groups to facilitate the transmission of both genetic and cultural information. The results
487 further entail a number of testable predictions concerning the relationships between climatic
488 variables, population density, mobility, and the complexity of archaeological assemblages.

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493 as suggesting themes for future research.

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641 **Table 1.** Partial Pearson product-moment correlations between population density (individuals per
 642 km²) and distance moved per annum (km), controlling for climatic variables. ‘Lower’ and ‘Upper’
 643 columns give the 95% confidence intervals of each correlation, based on 10,000 bootstrap samples.
 644 ‘All’ indicates that MAP, MAT, SDP, SDT, ET & NAGP are all controlled for.

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Control	Correlation (r)			df	p
	Mean	Lower	Upper		
None	-0.830	-0.913	-0.747	173	<0.001
SDT	-0.715	-0.787	-0.632	172	<0.001
SDP	-0.760	-0.818	-0.692	172	<0.001
SDT & SDP	-0.711	-0.783	-0.626	171	<0.001
All	-0.709	-0.783	-0.620	167	<0.001

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665 **Table 2.** Performance of the 12 models shown graphically in Figure 2. Delta values (or 'AICc change'),
 666 relative likelihoods (L_i) and relative probabilities (w_i) are calculated as per Burnham et al. (2011). K
 667 is the number of parameters calculated in fitting a model, and $LN(L)$ is the model's log-likelihood.

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Set	Model	K	LN(L)	AIC	AICc	Δ_i	L_i	w_i
1	1	10	0.000	-	-	-	-	-
1	2	8	-1.359	18.718	19.585	0.000	1.000	0.385
1	3	9	-0.374	18.749	19.840	0.254	0.881	0.339
1	4	9	-1.359	20.717	21.808	2.222	0.329	0.127
1	5	9	-5.000	28.000	29.091	9.505	0.009	0.003
1	6	9	-17.431	52.861	53.952	34.366	3.447E-08	1.327E-08
2	7	10	0.000	-	-	-	-	-
2	8	8	-18.974	53.948	54.815	35.230	2.238E-08	8.619E-09
2	9	9	-4.015	26.030	27.121	7.535	0.023	0.009
2	10	9	-3.872	25.744	26.835	7.249	0.027	0.010
2	11	9	-1.360	20.719	21.810	2.224	0.329	0.127
2	12	9	-14.917	47.834	48.925	29.339	4.256E-07	1.639E-07

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685 **Table 3.** Regression analyses of latitudinal subsets of the population density data on the standard
686 deviation in annual precipitation (SDP). ‘Lower’ and ‘Upper’ columns give the 95% confidence
687 intervals of the beta coefficient. Degrees of freedom are $n - 2$ in all cases.

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Subset	Beta			n	p
	Mean	Lower	Upper		
Tropical	0.290	0.027	0.526	61	0.023
Temperate	0.625	0.485	0.745	99	<.001
Polar	0.625	0.334	0.849	15	0.013
All	0.618	0.517	0.701	175	<.001

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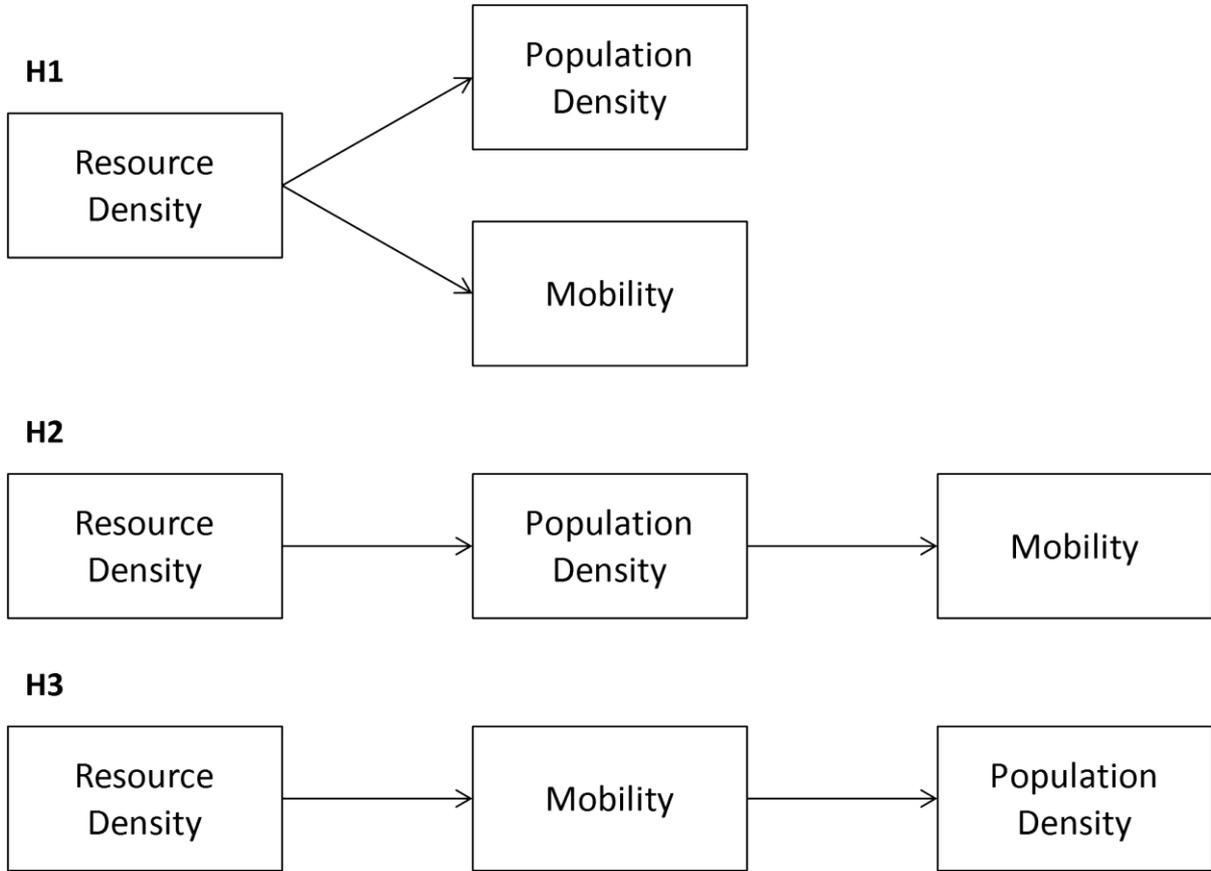
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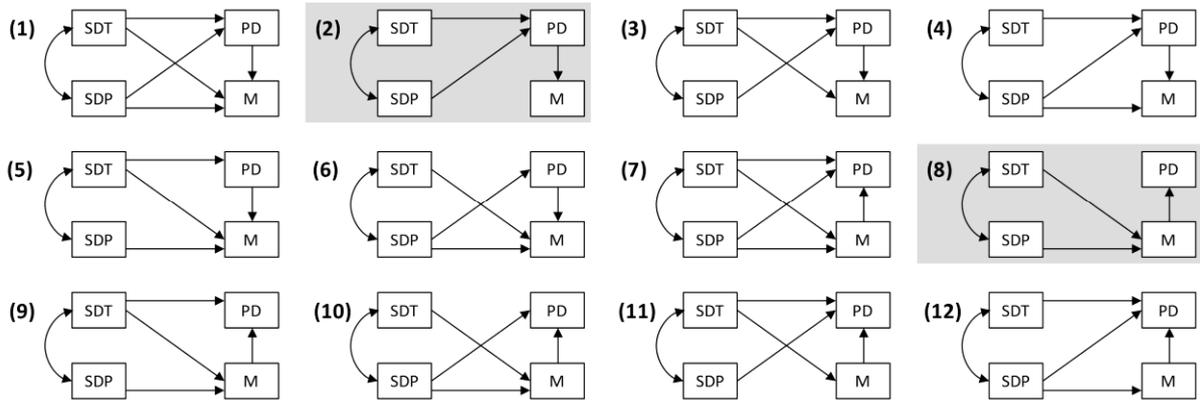
710 **Figures**

711 **Figure 1.** The three hypotheses shown as simple path diagrams. Arrows indicate hypothesized
712 causation.



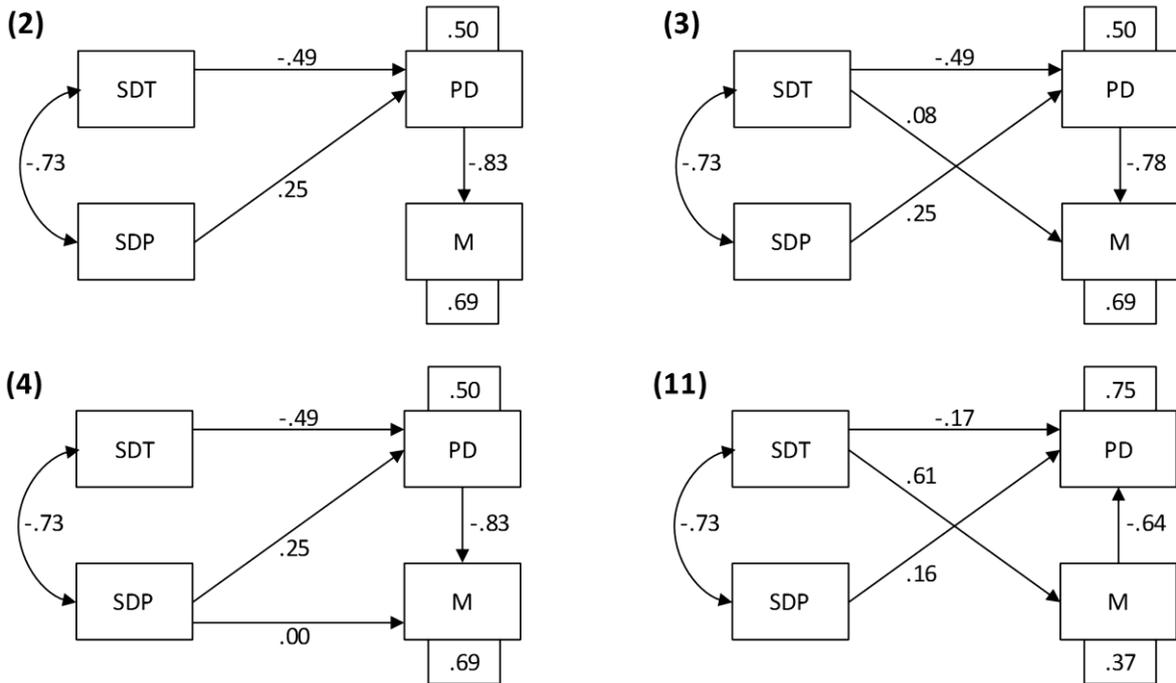
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725 **Figure 2.** The 12 possible causal models relating the two exogenous variables (SDT and SDP) to the
 726 two endogenous variables (PD and M). Models 1-6 represent variations of Hypothesis 2, whereas
 727 models 7-12 represent variations of Hypothesis 3. Model 2, the strongest embodiment of Hypothesis
 728 2, and Model 8, the strongest embodiment of Hypothesis 3, are highlighted with grey boxes. Arrows
 729 indicate causal paths for which path coefficients are calculated; curved double-headed arrows
 730 indicate correlations. SDT = standard deviation of temperature, SDP = standard deviation of
 731 precipitation, PD = population density, and M = mobility.



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749 **Figure 3.** The four best performing models from Figure 2 (see also the statistical output in Table 2).
 750 Calculated correlations and path coefficients are included. Labels in boxes are coefficients of
 751 determination (R^2) for the endogenous variables. For consistency, the models are numbered as they
 752 are in Figure 2 and Table 2.



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