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Human ecological niches and territories during the LGM in Europe derived from an application of eco-cultural niche modeling

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Abstract

We apply eco-cultural niche modeling (ECNM), an heuristic approach adapted from the biodiversity sciences, to identify habitable portions of the European territory for Upper Paleolithic hunter-gatherers during the Last Glacial Maximum (LGM), circumscribe potential geographic extents of the Solutrean and Epigravettian
technocomplexes, evaluate environmental and adaptive factors that influenced their distributions, and discuss this method’s potential to illuminate past human-environment interaction. Our ECNM approach employed the Genetic Algorithm for Rule-Set Prediction (GARP) and used as input a combination of archaeological and geographic data, in conjunction with high-resolution paleoclimatic simulations for this time frame. The archaeological data consist of geographic coordinates of sites dated by Accelerator Mass Spectrometry to the LGM and attributed to the Solutrean and Epigravettian technocomplexes. The areas predicted by ECNM consistently outline the northern boundary of human presence at 22,000–20,000 cal BP. This boundary is mainly determined by climatic constraints and corresponds well to known southern limits of periglacial environments and permafrost conditions during the LGM. Differences between predicted ecological niches and known ranges of the Solutrean and Epigravettian are interpreted as the Solutrean technocomplex being adapted to colder and more humid environments and as reflecting influences of ecological risk on geographic distributions of cultures.

**Introduction**

The idea of modeling past human-environmental interactions is by no means new. Researchers have used archaeological and environmental data sets, and diverse methods, to interpret prehistoric hunter-gatherer behavior in ecological contexts. Well-known European examples concern prehistoric population distributions during Oxygen Isotope Stages 2 and 3 (Gamble et al., 2004; Van Andel and Davies, 2003), as well as the resettlement of regions following severe climatic episodes (Gamble et al., 2005; Straus et
al., 2000). These studies were based on spatial distributions of radiometrically dated sites and generalized climatic reconstructions. Others have used a similar approach to estimate population size and kinetics (Bocquet-Appel and Demars, 2000; Bocquet-Appel et al., 2005). More detailed attempts to examine population distributions and human tolerances with respect to environmental variability also exist (Binford, 2001; Davies et al., 2003; d’Errico and Sanchez Goni, 2003; d’Errico et al., 2006). However, no agreement yet exists on how to best evaluate influences of environmental factors on prehistoric human populations and their responses to climatic variability.

One common limitation is the use of coarse-scale climatic data (i.e., simulations with resolutions of 3–5° in latitude and longitude) and imprecise chronological data (i.e., reliance on old conventional ages with large sigmas) that make evaluation of human responses to rapid-scale climatic variability, with adequate resolution, difficult. Another shortfall of previous studies is that they have incorporated environmental data into analyses only passively, such that these data are used as backdrops against which the archaeological record is interpreted. While these studies have obvious value, they are limited in their ability to evaluate prehistoric hunter-gatherer responses to the abrupt climatic and environmental changes of the last glacial period. The need for robust methods with which to evaluate more precisely how past human and animal populations responded to these changes is critical.

An important recent advance in the study of biological diversity has been the development of biocomputational architectures for predictive modeling of complex biodiversity phenomena (Guisan and Zimmermann, 2000; Soberón and Peterson, 2005). In particular, the ecological niche of a species can be defined as the range of
environmental conditions within which it can persist without immigrational subsidy (Grinnell, 1924; Hutchinson, 1957). Such tools can be used to predict species’ range expansion or contraction in response to real or simulated climatic changes (Peterson et al., 2002). They have considerable potential for reconstructing niches of past human populations and for illuminating the complex mechanisms that regulated the interactions between past hunter-gatherer populations and their environments, helping to shape cultural, genetic, and linguistic geography. This approach recently has been termed Eco-Cultural Niche Modeling (ECNM) (Banks et al., 2006). Our application of ECNM interactively integrates climatic, geographic, and archaeological data via a machine-learning genetic algorithm, described below. We argue that this approach is powerful and, when paired with high-resolution climatic simulations, allows us to overcome many limitations of previous studies and evaluate prehistoric human-environment interactions at regional scales.

Here, we apply ECNM to human populations at the Last Glacial Maximum (LGM) in Europe, a well-studied and dated climatic phase known to have had profound impacts on human populations, with three primary objectives: (1) to determine the limits of the potential human range during the LGM, (2) to define the eco-cultural niches of the two main archeological cultures present in Europe at that time (the Solutrean and Epigravettian technocomplexes), and (3) to identify environmental and cultural factors that shaped their geographic ranges.

Environmental and Cultural Context

The last glacial period was marked by dramatic and rapid climatic variability
(Dansgaard et al., 1993; Johnsen et al., 1992), with the LGM representing a unique suite of climatic conditions (Ditlevsen et al., 1996; Peryon et al., 1998). This period, centered on 21 kyr cal BP, is characterized by the maximum volume of the ice sheet over Scandinavia and northern Europe, along with cold and generally arid conditions in northern and Western Europe. The LGM archaeological record is characterized by a relatively small number of sites and large gaps in the archaeological record for many regions (cf. Soffer and Gamble, 1990; Straus, 2005; Street and Terberger, 1999). Such a pattern has been interpreted to be the result of the human abandonment of northern Europe and a contraction of the human range to southern regions that served as refugia. Such contraction and consequent demographic reduction is known to have produced a bottleneck in human genetic diversity (Barbujani et al., 1998:490; Torroni et al., 1998; Torroni et al., 2001).

In Western Europe, between ca. 22 kyr and 20 kyr cal BP, human groups responded to LGM environmental conditions by developing a suite of new technologies characterized by a variety of diagnostic projectile points and knives produced by bifacial retouch (Figure 1A), which define the Solutrean (Mortillet, 1873; Smith, 1966). Straus (2005) proposed that Solutrean populations employed more specialized subsistence systems, relative to earlier Upper Paleolithic technocomplexes, to exploit regions rich in game but under harsh climatic conditions.

In the regions of southeastern Europe, hunter-gatherers of the LGM produce a different lithic technology, termed the early Epigravettian (Laplace, 1964; Mussi, 2001), characterized by shouldered and backed projectile points produced by unifacial retouch (Figure 1B). Leaf-shaped points are rare and have been recovered from only a few sites in
northern Italy (Palma di Cesnola, 1990). Contrary to the Solutrean, which appears as a novel technology, the Epigravettian toolkit is interpreted as being derived from the preceding Gravettian technocomplex (Otte, 1990; Palma di Cesnola, 2001).

Materials and Methods

For ECNM, we employed a machine-learning genetic algorithm originally developed for determining the ecological niches of plant and animal species (Stockwell, 1999; Stockwell and Peters, 1999). This software application, termed the Genetic Algorithm for Rule-Set Prediction (GARP), has been applied to topics as diverse as habitat conservation, effects of climate change on species’ distributions, the geographic potential of species’ invasions, and anticipation of emerging disease transmission risk (Adjemian et al., 2006; Martinez-Meyer et al., 2004; Peterson et al., 2004; Sanchez-Cordero and Martinez-Meyer, 2000; Soberón and Peterson, 2004). It is available for download at http://www.lifemapper.org/desktopgarp/.

GARP requires as input the geographic coordinates where the target species has been observed and raster GIS data layers summarizing environmental variables that may be involved in limiting the geographic distribution of the species. In our application, the occurrence data were the geographic coordinates of radiometrically dated and culturally attributed archaeological sites. These archaeological data were obtained from a database, compiled by FdE and MV, that contains the geographic coordinates, recorded stratigraphic levels, associated cultural affiliations, and > 6000 radiometric ages from ca. 1300 archaeological sites in Europe.

The raster GIS data consisted of landscape attributes and high-resolution climatic
simulations for the LGM. The landscape variables included slope, aspect, elevation, and compound topographic index (a measure of tendency to pool water) from the Hydro-1K dataset (U.S. Geological Survey's Center for Earth Resources Observation and Science - http://edc.usgs.gov/products/elevation/gtopo30/hydro/index.html). Elevation was excluded from the jackknife manipulation models, described below, as generally recommended for that specific analysis (Pearson et al. 2007).

Typically, climatic simulations for specific periods of time are produced by forcing general circulation models (GCMs), which reconstruct past, present, and future climates globally typically at resolutions where grid squares measure 100–200 km on a side. For instance, the LGM and the mid-Holocene have been the focus of coordinated experiments in the framework of the Paleoclimate Modelling Intercomparison Project (Harrison et al., 2002; Joussaume and Taylor, 1995; PMIP, 2000). In an effort to use climatic data that approach the same scale of resolution as our geographic data, we use in the present study a regional climatic simulation with a grid box size over Europe of ~ 60 km on a side, which has been run at the Laboratoire des Sciences du Climat et de l’Environnement, Gif-sur-Yvette, France.

This high-resolution LGM atmospheric simulation followed the PMIP1 protocol (http://www-lsce.cea.fr/p mip), with sea-surface temperatures and sea ice cover as prescribed from the CLIMAP (1981) data set and the ice-sheets from the Peltier (1994) ICE-4G reconstruction. Atmospheric CO$_2$ concentration was lowered to 200 ppmv according to the ice-core record (Raynaud et al., 1993) and orbital parameters adjusted to 21,000 cal BP values (Berger, 1978). The results of this simulation have been compared to pollen-based climatic reconstructions, with fairly close agreement for summer and
annual mean temperatures but some underestimation of winter cooling and drying over Western Europe and the Mediterranean (Jost et al., 2005). From this simulation, we derived the following variables for input into GARP: warmest month temperature, coldest month temperature, mean annual temperature, and mean annual precipitation (Figure 2). The values of warmest and coldest months refer to the warmest/coldest month in a climatic cycle averaged over 10 yr of simulation.

In GARP, geographic locations of archaeological sites are resampled randomly by the algorithm to create training and test data sets. An iterative process of rule selection is then performed within the program’s functioning, in which an inferential tool is chosen from a suite of possibilities (e.g., logistic regression, bioclimatic rules) and applied to the training data to develop specific rules (Stockwell 1999). These rules maximize predictivity by using independent data to evaluate them. Predictive accuracy is evaluated based on the known presence data and a set of pseudoabsence points (i.e., points sampled randomly from among points across the region of study where the species has not yet been detected) (Stockwell 1999). This evaluation process is used to develop a rule-set that defines the distribution of a species in ecological space (i.e., an ecological niche) (Soberón and Peterson, 2005), which can be projected onto the landscape to predict a potential geographic distribution (Peterson, 2003:421; Stockwell, 1999; Stockwell and Peters, 1999). GARP has undergone extensive improvement and testing in recent years, including detailed sensitivity analyses (Peterson and Cohoon, 1999; Stockwell and Peterson, 2002a, 2002b; Anderson et al., 2002).

We applied GARP to archaeological sites dated by AMS to the LGM, in an effort to minimize the possibility of incorporating sites for which radiometric determinations
are underestimates of true ages, as has been shown to be common for older Upper Paleolithic technocomplexes that date to the temporal limits of radiocarbon methods (d’Errico and Sanchez-Goni, 2003; Zilhao and d’Errico, 1999). The lack of agreement between conventional and AMS ages is attributed by these authors to ineffective sample treatments and application of conventional $^{14}\text{C}$ counting methods to samples that fall near the limits of this dating method. While these factors should have a lesser impact concerning sites dated to the LGM, they still may be a source of error considering this period’s relatively narrow time window. Figure 3, presenting distributions of conventional and AMS ages from sites attributed to the Solutrean, indicates that such sources of error are present for ages during the LGM: conventional ages are slightly younger relative to AMS ages, suggesting that some underestimate the true age of their sites.

For this reason, the site samples used to create our Solutrean and Epigravettian ECNMs are composed primarily of sites radiometrically dated by AMS to the height of the LGM (defined here as $21 \pm 1$ kyr cal BP), and that contain diagnostic material assemblages associated with these technocomplexes (Table 1). AMS ages for sites assigned to the two technocomplexes of interest were calibrated using CALIB 5.0.2html (Reimer et al., 2004; Stuiver et al., 2005). The geographic coordinates of those sites that fell within our targeted time frame were used as occurrence points. We included 5 undated sites in Italy reliably attributed to the early Epigravettian of the LGM based on their stratigraphic contexts and diagnostic material assemblages to increase sample size for this technocomplex.

The other exception to our site selection protocol is the Solutrean site of Buraca
Escura. The conventional age from this site (Gif-4585) is very similar to the AMS ages from nearby Solutrean sites that, when calibrated, fall just outside the LGM timeframe. In all, geographic coordinates for 11 Epigravettian and 9 Solutrean sites were used as input to produce the GARP models.

We used the following specifications in GARP. Given the random walk nature of the method, we ran 1000 replicate runs, with a convergence limit of 0.01. Given the small size of the samples, we used ca. 80% of occurrence points for developing training rules in each analysis and reserved one point for model selection and one for evaluating model predictive ability. We then followed a protocol for selecting among the resulting models (Anderson et al., 2003), with omission error (i.e., failure to predict a known presence) measured based on the single reserved model selection point (see above), and models retained only when they were able to predict that single point (i.e., hard omission threshold of 0%). Commission error, conversely, is a measure of areas of absence that are incorrectly predicted present (Anderson et al., 2003:213). We followed recommendations of removing from consideration those 50% of models that show extreme values of proportional area predicted present. The resulting final ‘best-subset’ models ($N = 10$ for each technocomplex) were then summed to produce a best estimate of the potential geographic distribution for each technocomplex. This same procedure was used with all sites combined, regardless of cultural affiliation, to predict potential human range during the LGM.

Predictive models such as ECNMs are just that—predictions. As such, ECNMs must be tested for predictive accuracy before they can be interpreted. Given low occurrence data samples, we tested model predictions using the jackknife manipulation
proposed by Pearson et al. (2007), which is the only robust test for evaluating models based on small samples. Here, we used the single point set aside for evaluating model predictivity: if \( N \) occurrence points are available, \( N - 1 \) points are used to develop \( N \) jackknifed models. The success of each replicate model in predicting the single point that was omitted, relative to the proportional area predicted present, is then calculated using an extension to the cumulative binomial probability distribution (Pearson et al., 2007).

To evaluate whether the two technocomplexes reflect adaptations to two different ecological regimes, we compare their respective ecological niches. To do this, grid squares in the ecological matrix (temperature vs. precipitation) of each technocomplex that contained a predicted presence for all 10 GARP best-subset models were plotted against all the data in the study region from the climate simulation. Such plots were developed among mean annual precipitation, mean temperature, coldest month temperature, and warmest month temperature.

**Results**

The model produced using both Solutrean and Epigravettian sites identifies a clear northern boundary for potential human range during the LGM (Figure 4), which is also reproduced in the models for each separate technocomplex (Figure 5). This boundary follows the Loire valley in France, excludes the Massif Central, includes the Mediterranean regions of France, follows the southern limit of the Alps, and the northern limits of the Carpathian range (Figure 4).

The territories predicted for the Solutrean and the Epigravettian are presented individually in Figure 5. The Solutrean model predicts potential presence of human
groups associated with this culture in southwestern and southern France, northwestern portions of the Iberian Peninsula, the Ebro valley, and in disjunct areas of Cantabria, northwestern Italy, and the Balkans. The model for the Epigravettian predicts a potential presence of this culture in the Balkans, the Italian Peninsula excluding the most southerly regions, the Mediterranean regions of France as well as the Aude and Garonne corridors, and the Iberian Peninsula excluding its southern regions. That these models have high predictive power regarding potential human distributions is shown by the accuracy observed in the jackknife manipulations. The independent test point was correctly predicted in 7 of the 9 jackknife models for the Solutrean and in all 11 models for the Epigravettian, with associated probabilities of \( P = 0.00005 \) and \( P < 0.00001 \), respectively.

Although potential distributions predicted for these two technocomplexes show only minimal overlap geographically, conclusions of ecological differentiation are complex. These models are geographic projections of ecological niches defined by multiple environmental variables, so small differences between ecological niches can result in different potential geographic distributions when ecological differences correspond to environmental conditions present over large regions.

A Principal Component Analysis of the environmental variables indicated that overall environmental variability in the study area is satisfactorily explained (85%) by the first two components, which are most influenced by the different temperature variables associated with each technocomplex’s predicted distribution. Plotting the climatic variables’ values of the grid squares where all best subset models predicted potential presence against all of the available climatic data (Figure 6) showed that the ecological niches occupied by the two technocomplexes overlap broadly, with only slight
differences on the edges of their predicted niches. These differences indicate that the Solutrean technocomplex had the potential to occupy somewhat cooler and more humid environments than the Epigravettian. T-tests performed on these data matrices to compare the two technocomplexes were uniformly significant (p << 0.05), indicating that the Solutrean and Epigravettian niches are not drawn from the same population.

Discussion

The limits of the human range predicted by ECNM for the LGM (Figure 4) are arguably accurate. These limits are consistent with the known distribution of archaeological sites for this period (Bocquet-Appel et al., 2005; Demars, 1996; Soffer and Gamble, 1990). The only radiometrically dated site for our temporal range that seemingly contradicts our results is that of Wiesbaden-Igstadt (Street and Terberger, 1999), which has yielded 7 AMS ages from a single occupation level ranging from 19,320–17,210 BP. Street and Terberger (1999:267) think that these ages collectively represent the true age of the site but acknowledge, however, that uncertainties (e.g., contamination) could exist. When calibrated, two of these dates (UZ-3768 and OxA-7500) fall within our LGM window, but they are appreciably younger than the other calibrated dates from the same level suggesting that they underestimate the true age of the occupation. This interpretation is supported by the fact that when these two ages are averaged \( t_{\text{ave}} = 17,356 \pm 118 \) using the method described by Long and Rippeteau (1974), and compared to the next youngest age (OxA-7501), the null hypothesis of no difference is rejected \( t = 4.0143, P < 0.001 \). Because the younger and older ages from Wiesbaden-Igstadt cannot be considered to be drawn from the same statistical population,
and the older ages fall before the LGM when calibrated, we hesitate to accept that this site represents an LGM human occupation of the Central Rhineland. Wiesbaden-Igstadt probably dates to Dansgaard-Oeschger Interstadial 2 to which Shackleton et al. (2004:1515) assign an age of 19.62 ± 0.21 kyr BP.

One might argue that the site samples used in our study are not representative of human population distributions during the LGM. Some regions may have been only sporadically occupied leaving undetectable archaeological signatures. Such an argument is contradicted by statistical analyses (Bocquet-Appel et al., 2005) that convincingly show that the frequencies and distributions of recorded archaeological sites in Europe are representative, when considered with an appropriate taphonomic perspective, of prehistoric population distributions. Considering these arguments, the northern latitudinal limits of human occupation during the LGM, as indicated by our ECNM predictions, represent accurate estimates of the areas occupied by hunter-gatherers during this period.

Interestingly, the northern range of the GARP predictions corresponds to the southern boundaries of periglacial environments in Western and Central Europe (Huijzer and Vandenberghe, 1998; Lautridou and Sommé, 1981). The GARP limits in France follow closely those that separated regions characterized by continuous deep permafrost and continuous permafrost (Figure 7) (van Vliet-Lanoë et al., 2004). The limits predicted by the ECNM for southern France, Iberia, and Italy generally follow the boundary between continuous and discontinuous permafrost (Texier, 1996; vanVliet-Lanoë, 1996). Such correspondence strengthens arguments for the predictive power of our modeling approach since periglacial environments have low biomass, which may have prevented systematic utilization by prehistoric human groups.
The geographic distributions predicted by the ECNMs for the Solutrean and Epigravettian show only minimal overlap (Figure 5), suggesting that the two technocomplexes were to some degree adapted to different environments. Reconstructions of their ecological niches indicate that they overlap broadly, but that the Solutrean was able to exploit colder and more humid areas, corresponding to areas with permanent permafrost during the LGM. In contrast, the Epigravettian seems to have been more adapted to areas dominated by discontinuous permafrost and seasonal freezing. Neither technocomplex was adapted to the more southerly dry and relatively warmer Mediterranean environments during the LGM.

With respect to the correspondence between predicted and actual geographic ranges, the relatively restricted potential geographic distribution for the Solutrean corresponds well to this technocomplex’s archaeological distribution. In contrast, the Epigravettian ECNM prediction exhibits a potential range across southern France and Spain that is not corroborated archaeologically. This phenomenon is common (Peterson 2003) when species’ predicted ecological niches are projected across broad geographic areas: habitable areas are frequently predicted outside their observed range. Four factors are generally cited to account for such discordances: (1) limited dispersal, (2) speciation, (3) extinction on regional scales, and (4) competitive exclusion (Peterson, 2003:422–423). Limited dispersal refers to the inability of a species to occupy other regions due to physical mobility constraints, which does not seem to apply to the Epigravettian populations since coastal corridors were open and habitable during the LGM and would have allowed them to colonize western territories. Speciation also is not a factor since human populations that occupied Europe at the end of OIS 3 and during OIS 2 arguably
belonged to the same species. Likewise, regional extinction is not applicable, as it implies that Epigravettian populations were present in those regions before the LGM, but went extinct before its onset. The archaeological record demonstrates that such is not the case.

Of the factors proposed by Peterson (2003), only competition may explain the Epigravettian distributions. This hypothesis implies that Epigravettian populations could not occupy suitable regions of Western Europe, such as the northern Iberian Peninsula, because it would have been necessary to cross large areas occupied by competing human groups bearing a different cultural tradition, the Solutrean. This idea raises the question of why such competition would create a boundary between human groups, instead of resulting in occupation of the entire potentially exploitable geographic area by only one of them.

In the case of a biological species, the reasons that create a boundary between competing species are mainly ecological (MacArthur, 1972; Hutchinson, 1978). With humans, other factors can play roles in creating boundaries between groups. Contrary to most animal species, the carrying capacity of a human population is directly linked to its ability to maintain and transmit between generations not only a suitable technical system, but also a complex and dynamic set of social rules, cultural and religious values, systems of symbols, language, and ethnic identity. The geographic extent over which this heritage can be maintained may vary according to the nature of each human culture but is also highly dependent, particularly for hunter-gatherers, on ecological constraints. Nettle (1998) demonstrated convincingly that the geographic extent of linguistic entities increases in regions of high ecological risk, where ecological risk is defined as the amount of variation which people face in their food supply over time (seasonally or inter-
annually). Collard and Foley (2002) argued that cultural diversity decreases towards higher latitudes. Both studies attribute this pattern to the need to create long-distance social networks to increase the ability of human groups to survive in hostile environments. Limits to the expansion of such cultural and linguistic entities are thus arguably dictated by the need to maintain a degree of cultural and linguistic cohesiveness over these large ranges. The reconstructed ecological niches and their geographic projections for the two technocomplexes suggest that they occupied regions associated with different levels of ecological risk. Solutrean populations principally exploited regions characterized by colder and more humid conditions than those occupied by Epigravettian groups.

Good agreement exists between the predicted eco-cultural range for the Solutrean technocomplex and its actual archaeological distribution. We contend that the latter faced relatively high levels of ecological risk and consequently occupied as much of the potential geographic distribution allowed by their cultural adaptation as possible. In contrast, Epigravettian populations showed marked differences between potential and actual distributional areas. We suggest that Epigravettian groups faced lower levels of ecological risk and thus did not need to extend spatially as broadly.

Additionally, the potential geographic range predicted by the ECNM for the Epigravettian technocomplex during the LGM is restricted to a latitudinal band determined by geographic barriers (western Alps and Pyrenees)—given the narrowness of the European Mediterranean coastline, corridors between areas suitable for Epigravettian groups would have been restricted. Such irregular potential distributional areas would certainly have constituted major obstacles to maintenance of viable cultural
and linguistic networks across these regions. These restrictions were probably more effective between the Italian and Iberian Peninsulas than between the Italian Peninsula and the Balkans owing to the broad Adriatic plain created by low sea levels during the LGM (Antonioli et al., 2004; Lambeck et al., 2004). This difference certainly facilitated exchange between populations of the Italian Peninsula and southeastern Europe, as supported by Epigravettian affinities to LGM industries of the Balkans (Montet-White, 1996:121–122).

Conclusions

ECNM is an effective approach by which to characterize and quantify ecolinguistic niches associated with specific technocomplexes, and better understand how environmental factors influenced distributions of prehistoric human populations. Future research must focus on methods that can be used to evaluate more precisely the roles of competition and cultural cohesiveness in producing discords between predicted and actual ranges. The role of geography in constraining niches and creating cultural boundaries must be specifically targeted. Comparisons between the extent and location of niches occupied by technocomplexes under similar climatic and environmental conditions may allow us to better disentangle the influences of environmental and cultural factors. It also would be useful to apply these techniques to animal species of the LGM and other climatic episodes to evaluate whether or not their predicted ecological niches were associated with those of humans. Finally, the application of ECNM to historically documented hunter-gatherers, for which detailed cultural and environmental
data are available and cultural dynamics have been directly observed, would be instrumental in interpreting predictions based on archaeological and paleoclimatic data.

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References Cited


Stuiver, M., Reimer, P.J., Reimer, R.W., 2005. CALIB 5.0.2 [WWW program and documentation]. 14CHRONO Centre, Queen’s University Belfast, URL:http://www.calib.org.


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Figure 5: Eco-cultural niche models for the Solutrean and Epigravettian technocomplexes at the Last Glacial Maximum. For each technocomplex, grid squares with 1–5 of 10 models predicting the presence of an eco-cultural niche are indicated in grey, grid squares with 6–9 models in agreement are depicted in pink, and squares with all 10 models in agreement are indicated in red. Archaeological site locations (i.e. GARP occurrence points on which the models were based) are indicated by yellow circles.

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Figure 2
Figure 3
Figure 5
Figure 6