ANALYZING RADIOCARBON RESERVOIR OFFSETS THROUGH STABLE NITROGEN ISOTOPES AND BAYESIAN MODELING: A CASE STUDY USING PAIRED HUMAN AND FAUNAL REMAINS FROM THE CIS-BAIKAL REGION, SIBERIA

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ABSTRACT. Dietary offsets in radiocarbon dates are becoming increasingly interesting to researchers, not only because of their impact on the reliability of chronologies but also because of the possibilities for extracting further dietary information from the ¹⁴C data itself. This is the case with the cemeteries of the Cis-Baikal region being studied as part of the international Baikal-Hokkaido Archaeology Project set up to examine hunter-gatherer cultural dynamics in eastern Asia. Fortunately, to control for a freshwater reservoir offset, we were able to obtain a number of paired terrestrial herbivore and human material for ¹⁴C dating. This article tests the correspondence between stable isotope evidence and the offsets seen in ¹⁴C values and the implications for the analysis of the ¹⁴C measurements as "chronometric dates." This is an unusually well-documented example of freshwater reservoir offsets, providing an ideal case study to test different approaches to analyzing such offset information. Here, a purely Bayesian approach is compared with the more frequently applied linear regression analysis.

INTRODUCTION

The calibration of radiocarbon dates relies on having reliable information about the ¹⁴C measurements expected for samples of known age. For samples that get their carbon directly from the atmosphere, this is exactly what the atmospheric calibration curve provides. When the carbon in an organism comes from a different source, this is more problematic. For the oceans, a marine calibration curve gives an overall surface ocean average estimate (depleted in ¹⁴C by about 5%, or 400 yr), and deviations from this average are known for some regions in some time periods (primarily preindustrial, late-Holocene values, Stuiver and Braziunas 1998). For other reservoirs of carbon, such as lakes and rivers, the information is rather sparse and may have to be directly assessed as part of any dating study for which it is relevant.

There is a further level of complexity added when dealing with organisms, most notably humans, who draw their carbon from a number of different sources as a result of a mixed diet. However, the variations in diet are themselves interesting and provide extra information, which can help to disentangle the effects of the different reservoirs on the ¹⁴C content of samples.

Offsets in ¹⁴C dates due to dietary inputs are very important in some regions, where a dietary component from either marine (e.g. Dewar and Pfeiffer 2010; Naito et al. 2010; Ascough et al. 2012) or freshwater (e.g. Lanting and van der Plicht 1998; Cook et al. 2001; Shishlina et al. 2007, 2009, 2012; Olsen et al. 2010; Wood et al. 2013) systems is present. In some cases, these offsets are reasonably well correlated with measurements on stable isotope values or other indicators of the dietary system.

This article looks at a particular example of this type of problem, with which we can demonstrate some different approaches to data analysis. The sites involved in the study (Figure 1) are cemeteries in the Cis-Baikal region (Lokomotiv, Shamanka, Ust'-Ida, Kurma XI, and Khuzhir Nuge XIV) spanning the Neolithic and Early Bronze Age periods. The culture historical models for this region

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Proceedings of the Radiocarbon and Archaeology 7th International Symposium

Ghent, Belgium, April 2013 | Edited by Mark Van Strydonck, Philippe Crombé, and Guy De Mulder © 2014 by the Arizona Board of Regents on behalf of the University of Arizona



Figure 1 Location of middle Holocene cemeteries within the Cis-Baikal region of Siberia used in the study.

rely heavily on the ¹⁴C dating of the human remains (e.g. Mamonova and Sulerzhitskii 1989; Weber 1995; Weber et al. 2005, 2006). There is also very well-documented dietary variability at several levels: temporal, spatial, and individual (Weber and Goriunova 2011; Weber et al. 2011). The isotope ecology of Lake Baikal is unusual for a freshwater lake in having extremely variable δ^{13} C values in its primary producers, ranging from –30.5 to –9.0‰ (Kiyashko et al. 1998). This means that δ^{13} C alone cannot reliably identify the degree of consumption of foods from the lake. Stable nitrogen isotope (δ^{15} N) measurements largely reflect trophic level, and, as food chains are nearly always longer in aquatic systems (Minagawa and Wada 1984), they offer a means to investigate

human use of this resource. Both δ^{13} C and δ^{15} N values show that the humans from the region were certainly exploiting aquatic resources, from Lake Baikal itself as well as from the neighboring rivers (Katzenberg and Weber 1999; Weber et al. 2002, 2011; Katzenberg et al. 2009, 2010). However, given the only weak correlation between δ^{13} C and ¹⁴C offsets in this ecosystem, the focus in this paper is on δ^{15} N measurements.

The contribution of aquatic food inevitably varies between individuals as well as over time, and so we should expect a relationship between the stable isotope values for human bone collagen and the degree of depletion in ¹⁴C content in comparison with the atmosphere. This is a situation that is mirrored in many archaeological sites in different regions. The Cis-Baikal cemeteries have an advantage in that the burials frequently have associated with them a rich assemblage of artifacts, including objects made from the bones and teeth of terrestrial herbivores, whose ¹⁴C content should directly track that of the atmosphere. Assuming no significant curation of material, this allows us to obtain pairs of dates from burials (on human and herbivore bone/tooth collagen) which, in turn, enable us to measure directly the reservoir offset and thus study the relationship between the stable isotopes and the ¹⁴C offset in different ways.

THE RADIOCARBON DATA

For each of the graves dealt with in this study, we have measurements on human bone and animal teeth. In some instances, the human bone dates have been replicated and are here reported as a combined date (one of which fails a chi-squared test—see comments in Table 1 caption and results discussion). The underlying data are given in the online supplementary table file, and the details of the date combination and stable isotope data compilation are given in Schulting et al. (2014). Table 1 shows the main elements of the data relevant to the discussion here.

The offsets between the human and animal bone values are presented in Figure 2. There is in general a good linear relationship between the offset of the dates and their $\delta^{15}N$ values. The two most obvious outliers are data points 9 and 10 and in these specific cases there may be reason to exclude them from the analysis. For data point 9, the ¹⁴C dates on the human bone do not seem to be as reproducible as we would expect and the human burial represented by data point 10 is of an infant aged 2–4, likely still subject to a nursing effect responsible for its anomalously high $\delta^{15}N$ value (Waters-Rist et al. 2011).

LINEAR REGRESSION ANALYSIS

As a first-order approximation, we will assume a linear dependency between the stable isotope values for the humans and the offsets in the ¹⁴C dates of the human bones to those of the herbivores. Thus, if the parameters are defined as

- y_i the offset (¹⁴C yr) for the *i*th human in our data set
- x_{il} δ^{15} N value for the *i*th human in our data set
- x_{i2} δ^{13} C value for the *i*th human in our data set

the expected model for the offset is

$$y_i = \beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \varepsilon$$

where ε_i is a noise or error term specific to the sample, which might occur due to measurement error or due to the reservoir offset varying for reasons not reflected in the stable isotope measurements. We expect the ε_i terms to be normally distributed with a mean of zero. Given that the offsets y_i have a measurement uncertainty associated with them, we would expect ε_i to have a standard deviation to

Table 1 chi-squ other ir	I The data set on which th lared test on the human boundividuals are of post-wear	e analyses discussed her ne measurement (χ^2 test: ning age.	e are based. Fur $df = 1, T = 5.03$	ther details a (<i>cf.</i> 3.8)). **	tre given in Sc Data point 10	thulting et al. (2013) is on an infant with e	ınd in Table S levated δ ¹⁵ N c	 *Data-point lue to breastfee 	9 fails a ding; all
				luman samp	le		Herbivore san	nple	
Nr	Cemetery	Grave/Burial	Date (BP)	$\pm (1\sigma)$	ô ¹⁵ N (‰)	Species	Material	Date (BP)	$\pm (1\sigma)$
1	Lokomotiv	LOK_1980.022	6842	27	14.9	Marmot sp.	Incisor	6464	36
0	Lokomotiv	LOK_1981.024	6814	22	14.4	Marmot sp.	Incisor	6433	37
б	Lokomotiv	LOK_1988.038	6860	27	15.0	Marmot sp.	Incisor	6462	36
4	Lokomotiv-Raisovet	LOR_1980.003	6808	37	14.2	Marmot sp.	Incisor	6383	37
5	Lokomotiv-Raisovet	LOR_1997.011	7009	39	15.4	Marmot sp.	Incisor	6497	37
9	Shamanka II	SHA_2004.039	6859	26	14.1	Marmot sp.	Incisor	6336	35
٢	Shamanka II	SHA_2004.056	6935	38	15.4	Marmot sp.	Incisor	6313	36
8	Shamanka II	SHA_2005.059	6694	39	14.2	Marmot sp.	Incisor	6260	36
•6	Shamanka II	SHA_2008.104	6312	25	13.7	Marmot sp.	Incisor	6343	28
10^{**}	Ust' Ida	UID_1988.017	4887	29	15.5	Cervus elaphus	Canine	4762	30
11	Ust' Ida	UID_1988.018	4890	30	12.5	Cervus elaphus	Canine	4706	30
12	Ust' Ida	UID_1989.020.01	4874	24	13.1	Cervus elaphus	Canine	4608	30
13	Ust' Ida	UID_1989.021.01	4940	22	13.9	Cervus elaphus	Canine	4611	30
14	Ust' Ida	UID_1991.041	4876	33	12.8	Cervus elaphus	Canine	4695	32
15	Ust' Ida	UID_1994.052	4965	30	13.6	Cervus elaphus	Canine	4658	31
16	Ust' Ida	UID_1994.053.01	4798	23	12.4	Cervus elaphus	Canine	4644	31
17	Kurma XI	KUR_2002.001	3954	31	11.7	Cervus elaphus	Canine	3764	23
18	Kurma XI	KUR_2002.003	4207	33	15.5	Cervus elaphus	Canine	3806	31
19	Kurma XI	KUR_2002.004	4132	22	15.6	Cervus elaphus	Canine	3709	30
20	Kurma XI	KUR_2002.005	4204	31	15.2	Cervus elaphus	Canine	3755	31
21	Kurma XI	KUR_2002.006	4157	32	14.7	Cervus elaphus	Canine	3744	30
22	Kurma XI	KUR_2002.012	3979	33	11.4	Cervus elaphus	Canine	3779	30
23	Kurma XI	KUR_2002.013	4189	31	14.0	Cervus elaphus	Canine	3773	30
24	Khuzhir-Nuge XIV	K14_1998.036.01	3800	30	12.4	Cervus elaphus	Canine	3556	30
25	Khuzhir-Nuge XIV	K14_1998.037.01	3803	31	11.1	Cervus elaphus	Canine	3613	31
26	Khuzhir-Nuge XIV	$K14_{-1998.037.02}$	3728	22	11.7	Cervus elaphus	Canine	3610	30
27	Khuzhir-Nuge XIV	K14_1999.045	3829	29	14.4	Cervus elaphus	Canine	3528	28
28	Khuzhir-Nuge XIV	$K14_{-1999.049}$	3888	30	15.8	Cervus elaphus	Canine	3589	33
29	Khuzhir-Nuge XIV	$K14_{-}2000.077$	3790	21	10.4	Cervus elaphus	Canine	3824	30
30	Khuzhir-Nuge XIV	K14_2001.087	3833	30	12.8	Cervus elaphus	Canine	3697	31

C Bronk Ramsey et al.

be at least as large as this measurement uncertainty. Furthermore, in practice, we would also expect to see overdispersion because we know that variation in offset is unlikely to have a perfect linear relationship given variation in the end-point stable isotope values for the food sources and other physiological factors independent of diet.

Since in this particular case we have all of the values for each human (y_i, x_{i1}, x_{i2}) , a simple ANOVA linear regression is all that is required to obtain an estimate for β_0 , β_1 , and β_2 . The standard deviation of the residuals provides an estimate of the scatter in ε_i .



Figure 2 Plot showing the relationship between $\delta^{15}N$ and the offset seen between human and animal bones in the Baikal data set. Uncertainty in the offset is shown at 1σ . The two largest outliers from a clear linear relationship are data point 9 (triangle), where the human dates fail a χ^2 test, and data point 10 (diamond), which is from an infant burial.



Figure 3 Schematic graph of the models used in the linear regression analysis (black) and Bayesian analysis (gray). For the linear regression analysis the residual errors ε_i are assumed to be drawn from a Normal distribution with a mean of zero. For the Bayesian analysis the parameters z_i are modeled to be from a Normal distribution with a mean and standard deviation that are found by the model.

794 C Bronk Ramsey et al.

The application of linear regression to the Cis-Baikal data set is discussed in more detail in Schulting et al. (2014). In practice in this case, as in many others, it turns out that one parameter (the δ^{15} N value, x_{i1}) is sufficient to account for the variation in the offsets and there is no significant gain in including the second independent variable (the δ^{13} C value, x_{i2}) in the regression analysis. The regression model (Figure 3) then reduces to:

$$y_i = \beta_0 + \beta_1 x_{i1} + \varepsilon_i$$

It is convenient to calculate a derived quantity b, which is given by:

$$b = -\beta_0 / \beta_1$$

This gives the mean values of x_{i1} (δ^{15} N) for which we expect a zero offset in the ¹⁴C value. In other words, *b* is the δ^{15} N value we would expect for a human with no input to their diet from aquatic sources.

BAYESIAN MODEL

The Bayesian model that we propose follows the same logic as the linear regression. However, where it is potentially more powerful is that it can, in principle, be applied in cases where direct information on the offsets y_i in the ¹⁴C dates is lacking. Here, we use a single dependent variable model, but with enough data it would be possible to extend this to two or more variables.

We define one independent global parameter for the model (β_1), which is the change in offset for a shift of one in δ^{15} N value. The model (Figure 3) for the offsets becomes:

$$y_i = \beta_1 x_{i1} + z_i$$

We assume a uniform prior for y_i for the *n* skeletons, which from our knowledge of this particular basin is taken to be anywhere from 0 to 1000 ¹⁴C yr (Nomokova et al. 2013). Note that this parameter is not allowed to be negative, as there should not be ¹⁴C enrichment due to reservoir effects. The prior for β_1 is also uniform but for operational reasons constrained between 25 and 150 yr; these arbitrary end points are taken to be wide enough not to constrain the posterior (in such cases it may be useful to run models with different limits and choose ones that are wide enough not to constrain the posterior, but narrow enough to ensure fast model convergence); we choose to formally exclude zero in this case for reasons discussed below. The x_{i1} values are known. This leaves the parameters z_i : these are assumed to be drawn from a normal distribution with a mean μ and standard deviation σ both with uniform priors. Thus, overall we have

$$\beta_1 \sim U(25, 150)$$

$$y_i \sim U(0, 1000)$$

$$z_i \sim N(\mu, \sigma^2)$$

$$\mu, \sigma \sim \text{Uniform prior}$$

When compared to the linear regression model, we can see that β_1 still has the same meaning, and should be directly comparable; μ is equivalent to β_0 ; σ is equivalent to the overdispersion in ε_i . In all cases of these parameters, the Bayesian model produces a marginal posterior probability distribution rather than a point estimate. There are also other rather minor differences: for example, the prior for y_i is not allowed to go below zero—a constraint not applied in the linear regression model. As for the linear regression model, we can calculate a derived quantity b, which is given by

 $b = -\mu/\beta_1$

This gives us the $\delta^{15}N$ value we would expect for a human with no input to their diet from aquatic sources. Note that this value will be indeterminate if β_1 is zero, which is why the prior for this excludes that value.

Up to this stage, the Bayesian model assumes nothing about the relationship between different samples in the model. That is, the parameters y_i can be left unconstrained except by their prior. In this particular case, we will want to define that the animal bones associated with the burials are assumed to be the same age as the human bones. However, the approach used here could be applied much more widely to cases where the animal bones and human bones were, for example, just within the same sedimentary sequence. Here, we choose to apply it in a case where direct comparison with the linear regression model is possible.

IMPLEMENTATION OF THE MODEL IN OXCAL

This model can be conveniently implemented within OxCal (v 4.2.3) using the chronological query language (Bronk Ramsey 2009) with calibration against IntCal09 (Reimer et al. 2009). First, we set up the global β_1 parameter within the overall model:

Plot()
{
 beta_1 = U(25, 150);
 ...

Then, for each pair of samples we calibrate the animal bone (with no offset) and the human bone. For example, for the first sample we have

```
...
R_Date(6464,36);
y_1 = Delta_R(U(0,1000));
R_Date(6842,27);
...
```

with the offset for the first sample defined as being y_i with a uniform prior as defined above. We can then set up the parameters z_i and ensure that these are drawn from a Normal distribution. To do this, we put the parameters within a Gaussian phase model:

```
Sequence()
{
    s_1 = Sigma_Boundary();
    Phase()
    {
        z_1 = y_1-14.9*beta_1;
        z_2 = y_2-14.4*beta_1;
        ...
        z_30 = y_30-12.8*beta_1;
    };
    s_2 = Sigma_Boundary("S2");
};
sigma = (s_2-s_1)/2;
mu = (s_2+s_1)/2;
```

Next, we can calculate from the mean the parameter *b* that provides an estimate for the δ^{15} N of a terrestrial-feeding human:

...
b = -(s_2+s_1)/(2*beta_1);
b& = U(0,20,0.1);
...

The second line here does not in practice constrain b, but it ensures that the output probability distribution function is reported with a resolution of 0.1. Finally, for this particular model we wish to ensure that the human and animal bones are assumed to be of the same age. This can be done using the Combine statement wherever we have a pair of samples. More specifically, for the first pair we have

```
Combine("LOK_1980.022")
{
    R_Date(6464,36);
    y_1 = Delta_R(U(0,1000));
    R_Date(6842,27);
};
```

The full model is given in the online supplementary information.

RESULTS

We have applied both the linear regression analysis and Bayesian model to the full data set as well as to the data set without the two outliers (data points 9 and 10) identified above. The results of the four different models are shown in Table 2. The linear regression analysis provides single point estimates for the main parameters of interest, whereas the Bayesian analysis provides a probability distribution function for the parameters. In most cases, the outputs are approximately normally distributed. However, this is not the case for b, which has a skewed distribution (see Figure 4).

Table 2 The parameters derived from the linear regression and Bayesian models of the same data sets. In each case, Model 1 includes all of the data and Model 2 excludes the two outliers. Most of the parameters are explained in the text. For the linear regression model, the R^2 and adjusted R^2 values are the coefficients of determination giving some measure of significance of the fit and, in this case, indicating that for Model 1 with all the data about 43% of the variation is explained by the dependency on δ^{15} N and 70% when two outliers are removed. The *S* value is the standard deviation of the residuals. For parameters from the Bayesian model, we quote both a mean and standard deviation derived from the marginal posterior density functions.

Linear regression models			Bayesian models		
Parameter	Model 1	Model 2	Parameter	Model 1	Model 2
n	30	28	п	30	28
R^2	0.453	0.711			
Adjusted R^2	0.434	0.700			
β	-700	-788	μ	-662 ± 188	-775 ± 127
β	72.1	80.8	β	69.9 ± 13.7	79.6 ± 9.3
b	9.71	9.75	b	9.26 ± 1.00	9.66 ± 0.52
S	122	80.4	σ	105 ± 18	57.5 ± 16.6



which is the δ^{15} N we would expect for a human getting all of their food from purely terrestrial sources. Note that unlike many of the other parameters, with this model the marginal posterior density distribution for this parameter is skewed and not so well described by the mean and standard deviation.

The outputs from the two different approaches are very similar. Note that the underlying assumptions are not identical and so we would not expect the results to be identical. In particular, in the Bayesian model, the offset is constrained not to be less than zero, and the parameter σ measures the overdispersion of the data, whereas in the simple linear regression the *S* value is the standard deviation of the residuals. However, given that the average uncertainty in the offsets is only about 43 yr, *S* is mostly accounted for by overdispersion. We did also try a linear regression that also included variation with respect to δ^{13} C but, for Model 2, for example, this only increased adjusted R^2 from 0.700 to 0.702 and decreased *S* from 80.4 to 80.1. This shows that adding a second variable to this model does not improve it significantly (p = 0.288).

With the linear regression analysis, the offsets derived could be used to correct the ¹⁴C dates prior to calibration, and the scatter on the residuals added to the ¹⁴C date uncertainties (in quadrature) to account for the extra noise seen in the data, which is not accounted for by the variation in stable isotopes. For the Bayesian analysis, there are three different approaches that could be taken. First, one could leave the parameterization of the model as it is but include all human samples where there is no direct information on the ¹⁴C offset; this would allow the constrained data to inform the model more generally. Second, one could use the posteriors from this model as priors for a model with other human samples that one would expect to follow the same pattern, but for which we have no direct constraining information. Finally, one could use the central parameter estimates in the same way as one would for the linear regression model.

CONCLUSIONS

Bayesian modeling provides an alternative approach to dealing with offsets that are dependent on some observed variable such as δ^{15} N or δ^{13} C. Where we have the means to measure the offsets directly, as is the case here with the burials from Cis-Baikal, either method provides essentially similar results. The linear regression has the advantage of simplicity and that it generates diagnostic measures (R^2 and adjusted R^2 values). However, in cases where the offset cannot be directly measured, the Bayesian approach provides the methodology needed to include the offset analysis within a larger chronological Bayesian analysis.

798 C Bronk Ramsey et al.

Being able to build these ¹⁴C offsets directly into Bayesian chronological models allows us to make quantitative use of this information, and thus to improve their accuracy. Such improved accuracy is of particular importance in the Cis-Baikal region where the human dates play such an important role in understanding the history of the local cultures. This and the forthcoming study by Schulting et al. (2014) mark only the beginning of untangling this very complicated matter.

ACKNOWLEDGMENTS

This research was conducted for the Baikal-Hokkaido Archaeology Project funded as a Major Collaborative Research Initiative (MCRI) program of the Social Sciences and Humanities Research Council of Canada. We also acknowledge support for the Oxford Radiocarbon Accelerator Unit from the Natural Environment Research Council (UK) as part of the NERC Radiocarbon Facility.

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