

# The Effect of Ancient DNA Damage on Inferences of Demographic Histories

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The field of ancient DNA (aDNA) is casting new light on many evolutionary questions. However, problems associated with the postmortem instability of DNA may complicate the interpretation of aDNA data. For example, in population genetic studies, the inclusion of damaged DNA may inflate estimates of diversity. In this paper, we examine the effect of DNA damage on population genetic estimates of ancestral population size. We simulate data using standard coalescent simulations that include postmortem damage and show that estimates of effective population sizes are inflated around, or right after, the sampling time of the ancestral DNA sequences. This bias leads to estimates of increasing, and then decreasing, population sizes, as observed in several recently published studies. We reanalyze a recently published data set of DNA sequences from the Bison (*Bison bison*/*Bison priscus*) and show that the signal for a change in effective population size in this data set vanishes once the effects of putative damage are removed. Our results suggest that population genetic analyses of aDNA sequences, which do not accurately account for damage, should be interpreted with great caution.

## Introduction

Ever since genetic material from an extinct species was first analyzed in 1984 (Higuchi et al. 1984), interest in the field of ancient DNA (aDNA) has increased rapidly. Thanks to the enormous power of polymerase chain reaction (PCR) and the more recent developing of massive parallel sequencing, aDNA studies today include large-scale population genetic (Shapiro et al. 2004) as well as genomic projects (Noonan et al. 2005, 2006; Green et al. 2006; Poinar et al. 2006). This development will no doubt be of great importance for our understanding of general evolutionary processes. DNA retrieved from old remains, such as that of sediments stored under the Greenland ice cap (Willerslev et al. 2007), can literally open doors to lost worlds. By including ancient samples in population genetic studies, we can widen the time window through which we can make population genetic inferences. This allows us to trace events from which signals are no longer present in modern day populations. We may for instance be able to see past a recent bottleneck or understand when current patterns of diversity were established (Hofreiter et al. 2004). The aDNA studies allow us to track changes in genetic variation over time, which in turn can be correlated with for instance climate change or human impact (Shapiro et al. 2004). DNA from ancient specimens may also be used to test assumptions or models such as the molecular clock (Ho et al. 2005; Ho, Shapiro, et al. 2007) used in molecular evolution.

At the core of all aDNA work remains the postmortem instability of nucleic acid. Over time DNA is degraded and fragmented into smaller and smaller pieces, in the end leaving little or no amplifiable templates. As a consequence, PCR becomes very sensitive to contamination from other DNA sources. Apart from fragmentation, postmortem DNA damage can also lead to alteration of the individual bases that are present in the sequences generated post-PCR. Termed miscoding lesions, these can be explained by a combination of base degradation to derivatives that are sub-

sequently misreplicated by polymerase enzymes, plus the compounded effects of innate enzyme infidelity on typically low numbers of initial PCR templates (Hansen et al. 2001; Binladen et al. 2006). With regards to base degradation, although there has been some controversy as to exactly which base lesions are formed from such alterations, much evidence now points toward the deamination of cytosine (C) to uracil (U) as the dominant source of damage (Stiller et al. 2006; Briggs et al. 2007; Brotherton et al. 2007; Gilbert, Binladen, et al. 2007). During subsequent enzymatic replication (e.g., during PCR), these uracils are replaced with thymines (T), via an intermediate step where uracil leads to adenine incorporation opposite on the complementary DNA strand (Pääbo 1989). Ultimately, this process leads to the generation of C → T miscoding lesions. With regards to innate enzyme error, the principal problem is that, although error may be typically low, when PCR is initiated from very small numbers of templates, any error that occurs early during the reaction cycle might lead to a substantial number of descendent molecules containing the error (e.g., Hofreiter et al. 2001). In turn, when these are sequenced, the effect of this error may be severe. Due to both these problems therefore, the usefulness of aDNA studies often become critically dependent on precautions taken with respect to the status of the template DNA.

In recent years, population genetic theory has undergone a phase of rapid development. In particular, this progress has meant that coalescent theory now forms the basis of most population genetic methods. Furthermore, for studies including aDNA, or rapidly evolving organisms such as viruses, extensions to this theory now allow for inferences from data sets including temporally separated samples (Rodrigo et al. 1999). At the heart of these analyses lie inferences of population genetic parameters that influence the appearance of the genealogy of a locus. In this study, we investigate the effect of DNA damage, using a simple model of base degradation as an example, on inferred genealogical structure and its subsequent consequence for demographic inferences.

Miscoding lesions may impact the shape of inferred genealogies in 2 ways. First, false variation can cause misleading relationships among samples (Ho, Heupink, et al. 2007). Second, it may inflate the length of lineages in the tree connected to the leaf nodes. Here we focus on

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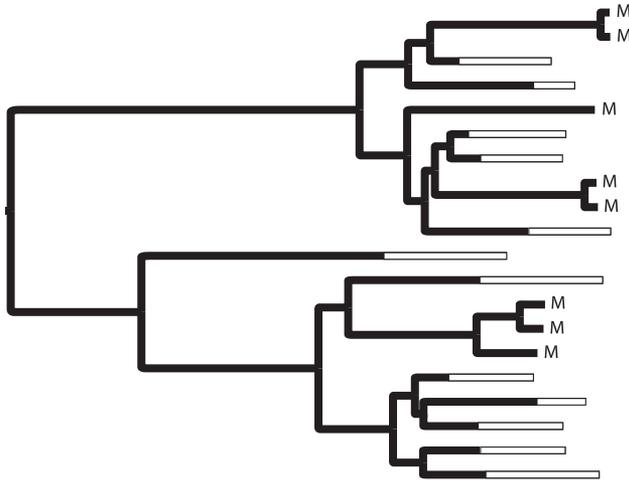


FIG. 1.—Hypothetical inferred genealogy from aDNA data (modern samples in the genealogy are labeled “M”). The filled branches (black) represent divergence due to true mutations, and the unfilled (white) branches represent artifactual divergence at the leaf nodes due to damage. As aDNA damage primarily arises as singletons in the data, primarily the branches leading to leaf nodes are extended by aDNA damage. The effect is to have longer branch lengths, and longer coalescence times, at the time around, or after, the sampling of the aDNA, leading to estimates of larger  $N_e$  at these times.

the latter of the 2 effects and illustrate how this problem may affect the inference of past population size change. In the standard coalescent model (e.g., Kingman 1982a, 1982b; Hudson 1983), coalescences occur at rate  $n(n-1)/4N_e$ , where  $n$  is the number of ancestral lineages and  $N_e$  is the effective population size. Genealogies of populations of large size are thus expected to exhibit relatively long coalescent times, whereas the opposite is expected for small populations. The predominant effect of miscoding lesions is to extend inferred branch lengths at around the time where the aDNA sequences were sampled. The artifactually elongated branch lengths will be interpreted as a reduction in coalescence rates and consequently also an increase in population size. In figure 1, the genealogy of a sample including both modern and ancient sequences is depicted. The example shows how genetic variation added to the sample as a result of damage accumulates on edges connecting ancient specimens to the rest of the tree. Simultaneously, both external branches leading to modern samples and internal branches of the genealogy are left largely unaffected by the false diversity. In the case of only aDNA, we would likewise predict artifactual inferences of population growth. However, for population samples including both damaged aDNA and modern DNA, we would expect coalescent-based methods to suggest demographic scenarios where an initial phase of growth is followed by a relatively recent decline, even though the actual population size was kept constant.

## Materials and Methods

To investigate this prediction further, we simulate serial samples of DNA sequences evolving under a constant population size and no recombination (most population

studies including ancient samples published to date are based on mitochondrial loci). We assume a mutation rate of  $10^{-7}$  mutations per site per year, occurring according to the Jukes and Cantor (1969) model of sequence evolution. However, our conclusions are not specific to any particular model of DNA sequence evolution. If appropriate, we use an exponential decay function at a rate of  $3.2 \times 10^{-7}$  miscoding lesions per site per year (Gilbert, Binladen, et al. 2007) to damage ancient samples with C  $\rightarrow$  T and G  $\rightarrow$  A transitions. Although such damage can arise due to both base degradation and enzyme sequencing infidelity, we simulate our rate of damage using values that relate to observed occurrences of base degradation only. Calculated from an exceptionally well-preserved, frozen (and thus relatively thermally young) mammoth bone, this rate is likely a conservative underestimate of the rate of base degradation in many ancient bone samples. Thus, we highlight that in a situation where all, or a subset, the aDNA sequences in a data set are generated without any form of replication, the damage rate used here likely represent a relatively modest number of miscoding lesions per sequence. In contrast, should replication of some form be applied to the data (including the generation of sequence from multiple PCRs or cloned sequences), damage may have less of an impact than that simulated here (cf. Hofreiter et al. 2001). The simulated damage corresponds to deamination of cytosine as the sole cause of DNA degradation, however, due to the complementary nature of the genetic code, and the fact that a change may have originated on either of the 2 strands, both types of transitions are simulated (Hansen et al. 2001). Finally, we run BEAST (Drummond and Rambaut 2007) under a Bayesian skyline model of population size change in order to infer the demographic history of the sample. The BEAST analyses extend over 10 million iterations, and genealogies and model parameters are sampled every 1,000 iterations. Results are visualized using R and Tracer, version 1.8 (Rambaut A, Drummond AJ [2007] Tracer v1.8, available from <http://beast.bio.ed.ac.uk/Tracer>). In all our simulations, we use samples consisting of a combination of 25 modern and 25 ancient samples, thus representing a total sample size of 50. Sequence length was set to 700 bp. For each scenario, we run 20 separate analyses. Using linear interpolation, we then estimate the population size every 5,000 years starting at present times and in the end present an average value across all 20 simulations plotted against time.

## Results

Figure 2a shows the results of a scenario in which the 25 ancient samples are of equal age set to 50,000 years. Moving from past toward present times, we see an initial phase where estimates suggest a constant population size. This phase probably represents coalescent events primarily involving internal and thus undamaged branches of the genealogy. Then, in line with expectations from arguments presented above, a phase of spurious but dramatic population growth follows as we move closer to the sampling time of the ancient sequences. A peak is reached before the actual age of the aDNA, likely corresponding to a burst in coalescent events involving damaged sequences. Such

events can only take place after the time point at which ancient samples were included to the genealogy. The growth phase is next replaced by a rather steep decline. The fact that we observe a relatively large population size still some time after having passed 50,000 years is probably partly reflected in the way the generalized skyline plots averages population size estimates across a number of coalescent events (Drummond et al. 2005). Finally, at present times, when coalescent events only involve undamaged contemporary DNA sequences, estimated population sizes are reduced to those used in the simulations.

We next simulate 20 new genealogies, this time without the addition of miscoding lesions to the ancient sequences. Figure 2b presents the results of these simulations, and this time we observe a flat line in good agreement with our constant-sized simulation settings. In order to imitate scenarios that resemble real data sets more closely, we also run simulations where aDNA ages are drawn randomly from a uniform distribution between 0 and 100,000 years. The results of these simulations, with and without damage, are shown in figure 2c and d, respectively. The general trend is similar to that of the previous simulations, although the false signal of population growth and decline is now prolonged as a result of the diversity in sample age. These simulations suggest that, unless stringent measures are taken to minimize their effects, miscoding lesions that can be expected to be present in sequences sampled from ancient specimens have the potential to create spurious signals of population size change. Furthermore, this signal has a characteristic shape where an initial apparent population growth is followed by a decline.

Next, in order to examine the effect of different damage rates on inferences of historical populations sizes, we perform simulations at 3 different damage rates ( $10^{-6}$ ,  $10^{-7}$ , and  $10^{-8}$  miscoding lesions per site per year). Although explicit damage rates have rarely been calculated in past studies, previous observations on base degradation alone have indicated that the rate of cytosine deamination is up to  $10\times$  lower in ancient frozen hair than frozen bone (reflecting a rate of  $10^{-8}$ ) (Gilbert, Tomsho, et al. 2007), whereas the higher rate of  $10^{-6}$  might be observed in extremely old nonfrozen materials or where significant levels of enzyme driven error are present. Again, we use 20 simulations for each rate and ancient sequences are of equal age (50,000 years). Figure 3 shows the estimated population size at the time of the sampling of the ancient sequences divided by the true population size (bias) plotted for the 3 damage rates. The results clearly show that the population size bias is affected by the damage rate and thus corroborates our finding that aDNA damage cause coalescent-based methods to overestimate past population sizes.

#### Reanalysis of Bison Data

Our analyses demonstrate that caution should be exercised when drawing conclusions from population genetic studies that include damaged sequences. Attempting to correlate the demographic history of species with for instance human activity or climate change might in some cases be premature and give misleading results. In the light of our

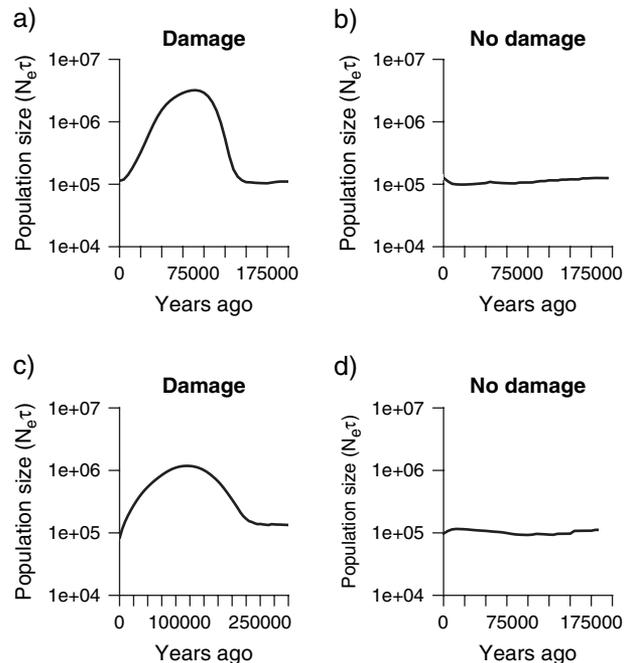


FIG. 2.—Demographic histories inferred by BEAST (Drummond and Rambaut 2007). The  $x$  axis is measured in years since present, and the  $y$  axis shows the product of the effective population size ( $N_e$ ) and generation time ( $\tau$ ). All simulated genealogies consist of 25 modern and 25 ancient sequences sampled from a population evolving under a constant population size (5,000 individuals). (a) All ancient sequences are sampled 50,000 years before present with damage as described in the main text. (b) Same as (a) but without damage. (c) The ages of the ancient sequences are drawn randomly from a uniform distribution ranging from 0 to 100,000 years, with damage as described in the main text. (d) Same as (c) but without damage.

findings, it might thus be worthwhile revisiting some of the published data sets in order to understand whether damage may have influenced previous results. To do this, we have chosen the largest yet published aDNA data set, on the steppe bison (*Bison bison/Bison priscus*) (Shapiro et al. 2004), to investigate whether the presence of reasonable amounts of miscoding lesions would affect the conclusions. When analyzed in its original form, without taking putative effects of damage into account, the data set shows a demographic history similar to those of our simulated genealogies containing miscoding lesions (Shapiro et al. 2004; Drummond et al. 2005). We remove all segregating sites that are due to transitions (153 out of 193) from the data set, thereby removing all, or almost all, false diversity that might be present. Reanalyzing the reduced data set in BEAST under the skyline model, we fail to see any signs of population size change (fig. 4a). Would all spurious diversity be due to base degradation and thus likely have arisen through the deamination of cytosine only, an outgroup could potentially have been used to determine the direction of transitions so that only  $C \rightarrow T$  and  $G \rightarrow A$  transitions could have been removed. However, as frequent and spurious  $T \rightarrow C$  and  $A \rightarrow G$  transitions have been observed in several aDNA data sets, generated on a wide range of substrates by a wide range of research groups (Gilbert et al. 2003), we do not adopt this approach here.

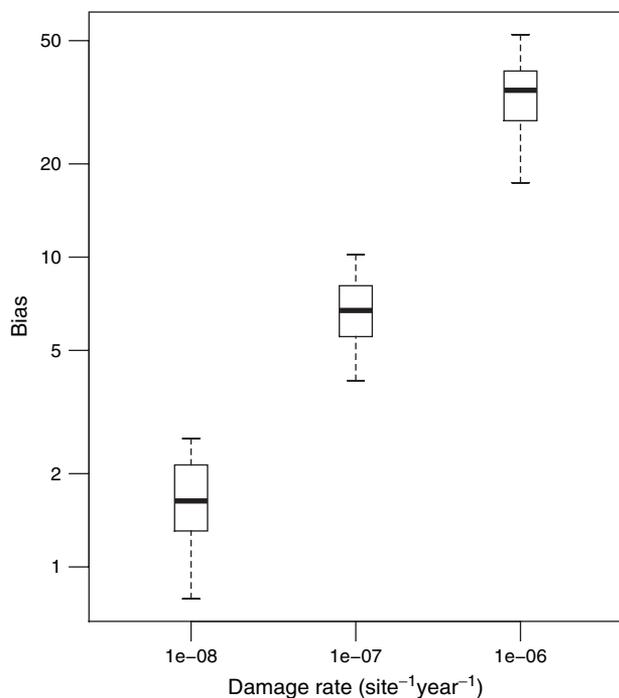


FIG. 3.—Bias in population size estimates as a result of DNA damage. The bias is the ratio of the estimated population size (in BEAST under a skyline model of population size change) at the time of the sampling of the ancient sequences (50,000 years ago) divided by the true population size used in the simulations. Simulations are performed at 3 different damage rates ( $10^{-6}$ ,  $10^{-7}$ , and  $10^{-8}$  miscoding lesions per site per year) and for each rate we generate and analyze 20 simulations. Whiskers in the box plot extend to the most extreme values of biases.

Clearly, the result of the analysis of the reduced data set might simply be due to a reduction in statistical accuracy and power due to the elimination of a large number of segregating sites (79%). Therefore in order to test if the reduced data set still holds enough statistical power to allow for the analysis, we perform several additional analyses. First, we produce 20 data sets where 79% of the segregating sites have been removed randomly without regards to the mutational type. The signal characteristic to the original set of sequences now reappears in these data sets (fig. 4*b*), indicating that the diverging results of the reduced and original data sets, respectively, are not explained by a simple lack of statistical power but more likely by the presence of miscoding lesions in the original bison data set—lesions that are not present in the reduced data set. Second, using an appropriate transition:transversion ratio, we simulate 3 different data sets consisting of 22 modern and 169 ancient, 700 bp long, sequences each (ages are uniformly distributed between 0 and 50,000 years), in the end harboring similar amounts of genetic variation as the real bison data set ( $\sim 200$  segregating sites). We then infer the demographic histories of these data sets prior to, and after, the removal of segregating transitions. Figure 5 displays the results of these simulations. Panel (a) shows the skyline plot of a sample taken from a population that recently increased 20-fold in size. As evident from panel (b), BEAST is able to recover the signal of population growth even after the removal of all segregating transitions from this data set. Panels (c) and (d)

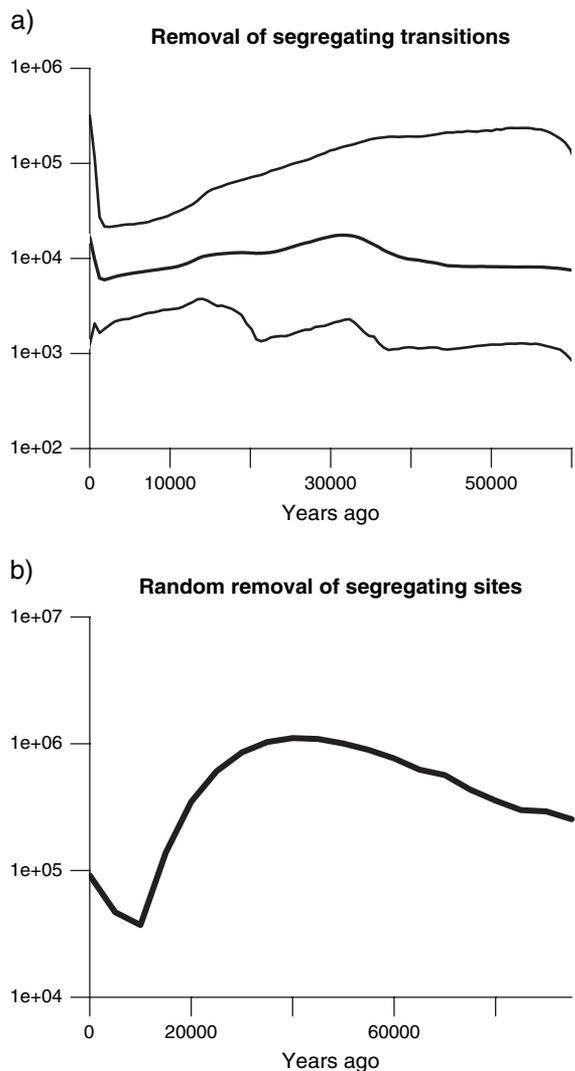


FIG. 4.—Revisiting a steppe bison data set. This figure depicts the results of 2 different Bayesian skyline reconstructions of the demographic history of a sample of 191 bison sequences used in a previously published study (Shapiro et al. 2004). (a) All segregating sites differing as a result of a transition were removed from the sample prior to analysis (median value and 95% confidence interval is shown). (b) A random set of segregating sites corresponding to a similar reduction in number of segregating sites as for panel (a) were removed prior to analysis. Panel (b) shows the mean demographic history inferred from 20 such randomly reduced data sets.

display the results of a scenario where the genealogy was simulated without population size change and again the shape of the skyline plots before (c) and after (d) removing all sites segregating as transitions both reflect the simulated demographic history well. In the third test, we once again simulate a sample from a population evolving under constant size; however, this time we also damage ancient sequences as described previously. Panel (e) displays how the population size of this simulated data is falsely inferred to have changed through time, while a signal consistent with the true demographic history is captured after the removal of segregating transitions prior to the analyses (panel f). The results of these simulations suggest that we are able to recover the demographic history of the bison using sites segregating as transversions only.

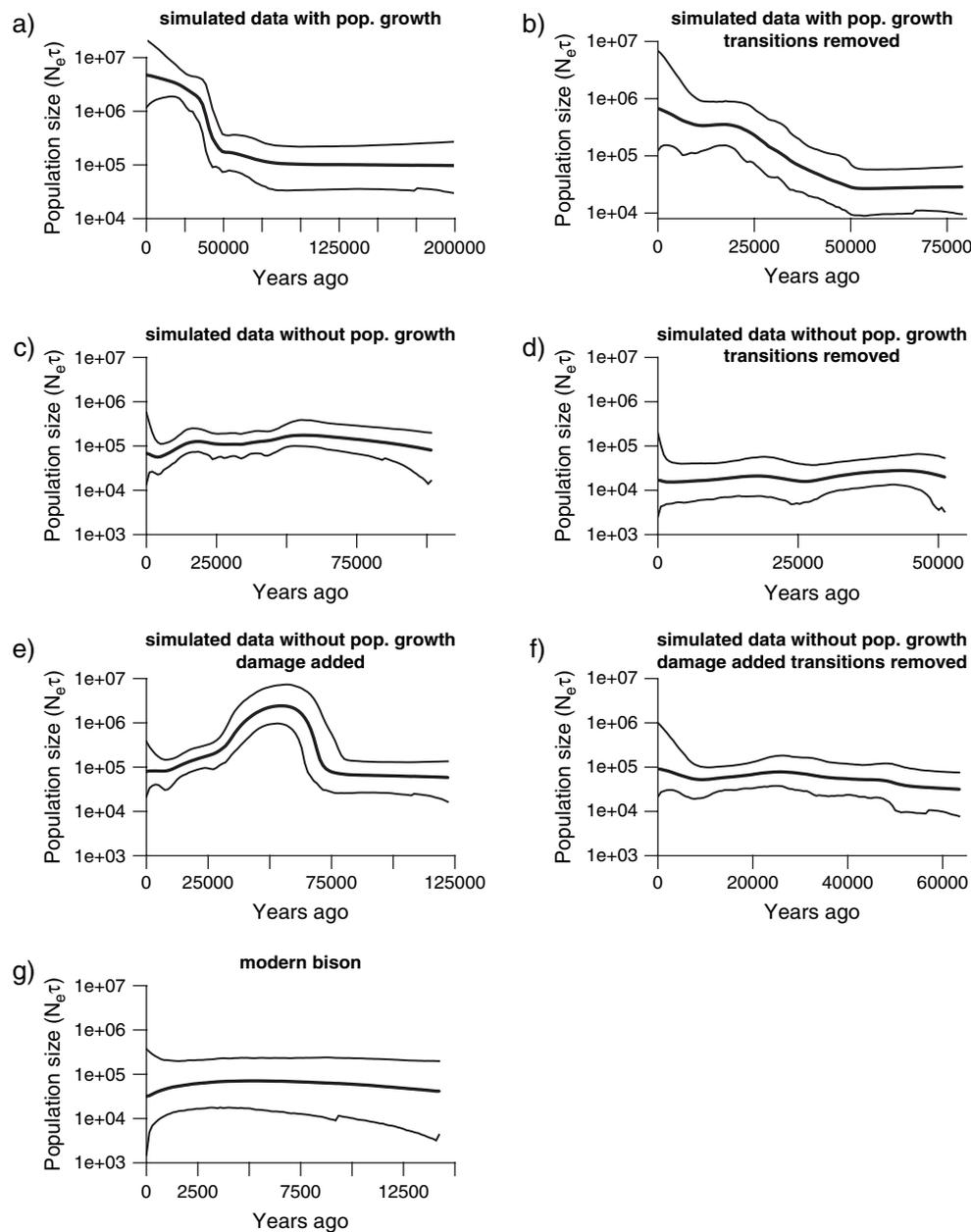


FIG. 5.—Panels (a–f) show the results of Bayesian skyline reconstructions of the demographic histories of 3 simulated data sets before and after sites segregating as transitions were removed. The data set depicted in panels (a) and (b) was generated under a model of population growth. Panel (a) shows the result of the analysis of the complete data set, as opposed to panel (b) where sites segregating as transitions were removed prior to analysis. Panels (c) and (d) display the inferred demographic histories of a data set that was generated under a model of no population size change, prior to, and after the removal of segregating transitions, respectively. Panels (e) and (f) present the demographic history of a data set simulated under no population size change, but where ancient sequences are affected by damage. In panel (e), a spurious signal of population size change is evident; a signal that is no longer present in panel (f) where segregating transitions were removed before the analysis. Finally, panel (g) presents the inferred demographic history of the steppe bison using modern samples only.

Finally, we infer the demographic history of the steppe bison using the 22 modern bison sequences included in the data set analyzed by Shapiro et al. (2004). These sequences should be unaffected by damage. As before we run BEAST under the Bayesian skyline model, this time with the mutation rate fixed at  $1.0 \times 10^{-7}$  mutations per site per year. Bearing in mind that, going back in time, these sequences may have coalesced before the time of the major demographic changes inferred in previous studies (Shapiro

et al. 2004; Drummond et al. 2005) (the diversity of the modern sequences may in other words not be informative for the time period of interest), it is still worth noting that DNA extracted from modern bison shows no signal of population size change (fig. 5, panel g). This result is expected if bison sequences that were collected from ancient specimens contain miscoding lesions. In conclusion, it seems likely that DNA damage has influenced the previous analysis of the demographic history of the steppe bison.

## Discussion

It is clear that DNA damage may be an important factor influencing population genetic studies. Although the preservation of some specimens used in aDNA studies is so good that damage rates may be negligible, this situation is rare in general and extremely rare in population studies. As such damage must be taken into account for most situations. There are several possible methods for eliminating the effects of DNA damage. A first is to treat aDNA extracts with the enzyme uracil-N-glycosylase (Pääbo 1989; Hofreiter et al. 2001; Johnson et al. 2007), which cleaves damaged templates at deaminated cytosines, preventing their subsequent PCR amplification. However, the downside to this is that the total template available can also drop significantly. A second is to only use aDNA sequences that have been obtained from multiple independent PCRs of the sample (Hofreiter et al. 2001; Willerslev and Cooper 2005). We should emphasize here that independent cloning from a single PCR is not sufficient evidence because of the possibility of PCR drift—the random increase in one haplotype over another due to stochastic drift in the early cycles of the PCR (e.g., Wagner et al. 1994). A third method for accounting for aDNA damage is to incorporate it into the statistical method of analysis. Recent progress has been made on this in the context of the BEAST/BEAUTI analytical package (Drummond and Rambaut 2007) that allows the incorporation of parameter,  $\delta$ , which models DNA damage (Ho, Heupink, et al. 2007). However, so far this model fails to properly accommodate for a time dependency of DNA decay. Inferences under such a model will thus still lead to biases when aDNA sequences have more damage than contemporary sequences and/or if there is any tendency toward older aDNA sequences having more damage than younger aDNA sequences. With this paper, we would like to encourage more research into statistical methods that could account for (and detect) aDNA in population genetic studies.

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