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Identifying a living great-grandson of the Lakota Sioux leader Tatanka Iyotake (Sitting Bull)

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A great-grandson of the legendary Lakota Sioux leader Sitting Bull (Tatanka Iyotake), Ernie LaPointe, wished to have their familial relationship confirmed via genetic analysis, in part, to help settle concerns over Sitting Bull’s final resting place. To address Ernie LaPointe’s claim of familial relationship, we obtained minor amounts of genomic data from a small piece of hair from Sitting Bull’s scalp lock, which was repatriated in 2007. We then compared these data to genome-wide data from LaPointe and other Lakota Sioux using a new probabilistic approach and concluded that Ernie LaPointe is Sitting Bull’s great-grandson. To our knowledge, this is the first published example of a familial relationship between contemporary and a historical individual that has been confirmed using such limited amounts of ancient DNA across such distant relatives. Hence, this study opens the possibility for broadening genealogical research, even when only minor amounts of ancient genetic material are accessible.

INTRODUCTION

Tatanka Iyotake, also known as Sitting Bull, was leader of the Hunkpapa Lakota Sioux who lived from 1831 to 1890 (1). For many years, he led his people in a fight against the policies of the United States for Native Americans, and he is particularly well known for his victory over General Custer in 1876 in the Battle of the Little Big Horn, also known as the Battle of Greasy Grass. Fourteen years after this battle, Sitting Bull was shot and killed by the Indian Police while they were attempting to arrest him (2). He was buried at Fort Yates in North Dakota on what was then known as the Standing Rock Agency, now the Standing Rock Indian Reservation. However, whether his remains are still there is a matter of debate. Some part of Lakota oral history says that his supporters moved his remains to an unknown location in Canada shortly after his burial (3, 4). In 1953, relatives of Sitting Bull engaged with a mortician to open the Fort Yates gravesite, and it is presumed that they moved his skeletal remains to a new gravesite in Mobridge, South Dakota. Today, two separate official gravesites exist for Sitting Bull, one at Fort Yates and the other at Mobridge. Both sites receive visitors wishing to pay their respects. Ernie LaPointe believes his relatives moved Sitting Bull’s remains to Mobridge and has concerns about the possible commercialization and care of the gravesite. To have the right to determine the fate of the gravesite, he needs to prove that he is a relative of Sitting Bull. The familial relationship between LaPointe and Sitting Bull has until now been based on birth and death certificates, a family tree, and a review of historical records (5). Genetic evidence would constitute an additional line of evidence and thus strengthen his documentation.

Fig. 1. Sitting Bull’s scalp lock. The lock that had been taken by H. Deebled and later brought to the Smithsonian Institution. Catalog EL00226, courtesy Department of Anthropology, Smithsonian Institution.
RESULTS

Sequencing the hair sample and assessing authenticity

First, we extracted DNA from Sitting Bull’s hair, built indexed Illumina sequencing libraries, and sequenced them on the Illumina HiSeq 2000/2500 platforms. After mapping, we removed duplicates and filtered away bases with base quality below 20 and reads with mapping quality below 25, which left us with sequencing data covering 0.8% of the genome with a depth of coverage of 2.7, or equivalently a mean sequencing coverage of 0.02×. Next, to assess the authenticity of the sample, we estimated base specific error rates of the quality-filtered sequencing data and found markedly higher error rates at SNP loci with transitions (fig. S1), which is typical of ancient DNA due to postmortem degradation (22). The remaining error rates were all below 0.001. Thus, the results of this analysis both corroborate the authenticity of the hair sample and suggest that most of the errors in the data can be avoided by removing transitions, which we did before performing all subsequent analyses. The authenticity of the sample is also supported by results from multidimensional scaling (MDS) analyses based on the sequencing data from Sitting Bull combined with SNP chip data from worldwide populations included in the Human Genome Diversity Panel (HGDP) (23). The MDS plot obtained for joint dataset clearly shows that the sample from the hair lock clusters with Native American individuals (Fig. 2) as would be expected since the hair sample is from Sitting Bull.

Assessing the familial relationship between Sitting Bull and Ernie LaPointe using a new method

To be able to assess the familial relationship between Sitting Bull and Ernie LaPointe, we genotyped 13 unrelated Lakota Sioux individuals, including Ernie LaPointe, using the extensive Illumina Omni 5M chip. Unfortunately, the sequencing data obtained from the hair sample were so limited in amount that the overlap between this and the chip data from Ernie LaPointe and the other Lakota Sioux individuals is restricted to 2259 polymorphic loci after excluding data from loci with transitions. This is not sufficient to perform reliable relatedness estimation for distantly related individuals using existing methods. Therefore, to investigate whether Ernie LaPointe can be genetically identified as the great-grandson of Sitting Bull, we developed a new probabilistic method tailored to the situation where one
has limited sequencing data from a historical individual and SNP chip genotype data from a present-day individual.

Briefly, the method is a maximum likelihood method that estimates a parameter, \( F_{\text{rel}} \), whose value depends on how the present-day individual is related to the historical individual. More specifically, \( F_{\text{rel}} \) is the probability that a randomly drawn allele from the historical individual, is identical by descent (IBD) with one of the present-day individuals’ two alleles in a SNP locus. Hence, if the two individuals are unrelated, then \( F_{\text{rel}} \) is 0, and if they are related, then the expected value of \( F_{\text{rel}} \) is 0.5, where \( m \) is the number of meioses between the two samples. In particular, \( F_{\text{rel}} \) is expected to be 0.5 \( \times 0.125 \) if the present-day individual is the great-grandson of the historical individual. Thus, estimating \( F_{\text{rel}} \) for Sitting Bull and Ernie LaPointe has the potential to reveal their true relationship.

The new method for inferring \( F_{\text{rel}} \) is based on the observation that for any given SNP locus, we can write the likelihood of \( F_{\text{rel}} \) as

\[
\Pr(SB | E, F_{\text{rel}}) = \Pr(\text{IBD} | E, F_{\text{rel}}) \Pr(SB | \text{IBD}, E, f, F_{\text{rel}}) + \Pr(\neg \text{IBD} | E, F_{\text{rel}}) \Pr(SB | \neg \text{IBD}, E, f, F_{\text{rel}})
\]

\[
= F_{\text{rel}} \Pr(SB | \text{IBD}, E) + (1 - F_{\text{rel}}) \Pr(SB | \neg \text{IBD}, f)
\]

where \( SB \) is a randomly drawn allele from the historical individual of interest (in this study, Sitting Bull) with possible alleles “0” and “1,” \( E \) is the genotype of the present-day individual of interest (in this study, Ernie LaPointe) with possible values of 0, 1, and 2, counting the number of 1 alleles, \( f \) is the frequency of the 1 allele in the population, IBD means that \( SB \) is IBD with one of the alleles that make up the genotype \( E \), and \( \neg IB \) means the opposite. More specifically, we can write

\[
\Pr(SB | \text{IBD}, E) = \begin{cases} 
1, & E = 0 \land SB = 0 \\
0, & E = 0 \land SB = 1 \\
0/2, & E = 1 \land SB = 0 \\
0/2, & E = 1 \land SB = 1 \\
0, & E = 2 \land SB = 0 \\
1, & E = 2 \land SB = 1
\end{cases}
\]

and

\[
\Pr(SB | \neg \text{IBD}, f) = \begin{cases} 
1 - f, & SB = 0 \\
f, & SB = 1
\end{cases}
\]

Assuming independence between SNP loci (i.e., no linkage disequilibrium), this means that the log likelihood of \( F_{\text{rel}} \) for all loci is equal to the sum of the log of the per locus log likelihoods, i.e., log of \( \Pr(SB | E, f, F_{\text{rel}}) \) from above. We obtain an estimate of \( F_{\text{rel}} \) using numerical optimization to identify the value of \( F_{\text{rel}} \) that maximizes this likelihood function.

Note that this method is based on two key assumptions (i) that the two individuals are from the same population and unadmixed and (ii) that allele frequencies, \( f \), from the relevant population can be obtained. In this study, we had access to genetic data from 13 unrelated Lakota Sioux individuals, of which 5, including Ernie LaPointe, are unadmixed according to an analysis performed with the genetic ancestry clustering tool ADMIXTURE (fig. S2). The remaining eight were inferred to have some degree of European ancestry in addition to their Lakota Sioux ancestry and thus be admixed. Consequently, we could apply the method to the data from Sitting Bull and Ernie LaPointe, but we only had five individuals to base allele frequency estimates on including Ernie LaPointe. Using these allele frequency estimates, the method provided an \( F_{\text{rel}} \) estimate of 0.1143 (table S2 and fig. S3), which is close to what we would expect if Ernie LaPointe is Sitting Bull’s great-grandson if the population allele frequencies are known. For comparison, the \( F_{\text{rel}} \) estimates for the other four unadmixed Lakota Sioux individuals are all below 0.002, suggesting that the \( F_{\text{rel}} \) estimates are not in general inflated among Lakota Sioux (table S2).

**Investigating what can be concluded from the obtained estimate**

To more properly assess what can be concluded from the \( F_{\text{rel}} \) estimate obtained for Sitting Bull and Ernie LaPointe, we performed simulations of data similar to the real data. More specifically, we simulated 100,000 datasets each consisting of low-depth sequencing data from a historical individual, SNP chip genotype data from four reference individuals, and two individuals for which we later estimated \( F_{\text{rel}} \): one, \( E_{\text{unrelated}} \), that is unrelated to the historical individual and one, \( E_{\text{great-grandchild}} \), that is the great-grandchild of the historical individual. Each of these dataset were simulated to have the same low number of SNP loci that are polymorphic among the four reference individuals and \( E_{\text{unrelated}} \) or \( E_{\text{great-grandchild}} (n = 2259) \) (Fig. 3). The results of applying the new method to these simulated datasets suggest that with the limited data at hand, an estimate of 0.1143 or higher can, in principle, occur even if Ernie LaPointe and Sitting Bull were not related but that this is unlikely to occur \((P < 0.00389)\). Thus, the results from the real data combined with the simulation results show strong support for Ernie LaPointe being Sitting Bull’s great-grandson.

Notably, assuming the simulations are sufficiently realistic, the simulation results also show that the method has 97.0% power to detect a great-grandchild if one uses a null model of the individuals being unrelated and a 0.05 significance threshold, making it a fairly powerful method for such distant relatives given the low amount of data available. Last, the simulation results show that the estimates provided by the method are biased when applied to the simulated data; it does not, on average, provide estimates that are equal to the expected value of \( F_{\text{rel}} \) (Fig. 3). When we reran the analyses for the same data using the true allele frequencies instead of frequencies estimated from only five individuals for the estimation of \( F_{\text{rel}} \), the bias was not present for the simulated great-grandchildren and markedly reduced for the unrelated individuals (fig. S4). In addition, the power to detect a great-grandchild is even larger (99.6%). This suggests that the bias is, to a large extent, caused by the fact that the sample size used to estimate allele frequencies is small and that the method would perform even better if more reference individuals were available.

**DISCUSSION**

This study first and foremost provides genetic support for Ernie LaPointe being the great-grandson of Sitting Bull, supporting the claim that he and his sisters were the rightful recipients of the repatriated items from the Smithsonian Institution. Second, it introduces a new approach to connecting the past to the present when very limited genetic data are available, as was the case in this study. The new approach requires data from at least a few individuals from the same population as the individuals of interest. It also requires that none of the individuals included in the analysis are admixed or inbred, since it relies on having representative allele frequencies. Last, it relies on simulations for assessment of the certainty of the results and thus requires that data similar to that of the real data can be simulated. However, when these requirements are met the new approach is simple and easy to apply. Hence, the new
The Sitting Bull hair sample

Ancient sample processing was performed in laboratories dedicated to ancient human remains (Lundbeck Foundation GeoGenetics Centre, Copenhagen, Denmark), following strict procedures to avoid contamination by modern or amplified DNA. We extracted DNA from a small hair lock following a protocol described by Rasmussen et al. (25). Three Illumina sequencing libraries were built from the DNA extract, using the procedure from Meyer and Kircher (26) with slight modifications (27). Each library was split in two halves and amplified in two steps (27) for 12 + 12 or 10 + 8 cycles, using AmpliTaq Gold (Life Technologies) and 6–base pair indexed primers. The final products were pooled and sequenced over one lane on the Illumina HiSeq 2000 platform, 93 single-read mode and one lane on the Illumina HiSeq2500 platform, 100 single-read rapid mode, at the Danish National High-Throughput DNA Sequencing Centre (Copenhagen, Denmark). The sequencing output was converted to fastq format using CASAVA, and reads with a length below 25 were discarded. The remaining reads were then trimmed and merged using AdapterRemoval (28) and mapped to hg19 (assembly hsbuilt37.1) using bwa (aln) (0.6.2) (29) with seed disabled (−l 1000). After mapping, we removed duplicates using SAMtools rmdup, and before all analyses of the sequencing data from the Sitting Bull sample, we filtered away all reads with mapping quality below 25 and bases with base quality below 20. The resulting data have a mean depth of 0.02×, and we refer to this as the quality-filtered sequencing data from Sitting Bull below.

SNP chip data from Lakota Sioux individuals

Saliva samples from 14 Lakota Sioux individuals, including Ernie LaPointe, were collected using the Oragene OG–300 DNA Self-Collection Kit (Genotek). We extracted DNA following the prepIT L2P protocol (Genotek). DNA extracts were processed for genotyping on the extensive Illumina Omniv5-Quad-Exome BeadChip at AROS Applied Biotechnology (Aarhus, Denmark) to maximize the overlap with the Sitting Bull sequencing data. This led to data from 4,334,816 SNP loci. Among the 14 individuals, two were self-reported to be half-siblings, and one of these was removed from the dataset before all analyses.
Other samples
For several of the analyses, we also used genotype data from the 1000 Genomes project and HGDP and sequencing data from individual NA12778 from the 2013 release of 1000 Genomes Project (30).

Dataset used for error rate estimation
We used the quality-filtered sequencing data from the Sitting Bull hair sample and used high-quality sequencing data for the genome of individual NA12778 from the 2013 release of 1000 Genomes Project as the high-quality genome (30). To determine the ancestral alleles and, thus, the derived alleles, we used the outgroup species chimpanzee as is standard in human studies. In particular, we used the multiway alignment that includes both chimpanzee and human (panTro2 from the hg19 multiz46).

Dataset used for MDS analysis
We used the HGDP data provided with the bammds software tool (23, 31), which contain data from 644,074 SNP loci for a range of populations from across the world. From these data, we removed all transition SNP loci, since these are much more error prone (fig. S1). This left us with genotype data from 120,643 SNP loci. The quality-filtered sequencing data from the Sitting Bull hair sample overlapped 654 of these loci, and these were the reads that were included in the analysis along with the genotype data. The extraction of these reads was performed by the bammds software tool that we used for the MDS analysis.

Dataset used for ADMIXTURE analyses
We merged the Lakota Sioux SNP chip data with genomic data for 40 individuals from the CEU 1000 Genomes population keeping only the SNPs that overlap and are consistent between the two datasets. After merging, we removed SNPs with minor allele frequency below 5%, SNPs with any missingness and SNPs in high linkage disequilibrium (using the PLINK option --indep-pairwise 100 10 0.5) leaving us with data from 363,519 SNP loci.

Dataset used for estimation of $F_{rel}$
We combined the quality-filtered sequencing read data from the Sitting Bull hair sample with the SNP chip data from the 13 not closely related Lakota Sioux individuals. When doing so, we first discarded all SNPs that did not overlap with any read data from Sitting Bull. This left us with data from 49,347 SNP loci. Subsequently, we removed all SNP loci with transitions, since these are much more error prone (fig. S1), which left us with 10,142 loci. Then, for each of the remaining loci, we sampled a single read from Sitting Bull, and if the read had one of the two alleles present in the SNP chip data at that locus, we included the locus in the analyses; if not, we discarded the locus, leaving us with 10,041 SNP loci. Last, we removed 8 of the 13 Lakota Sioux individuals because they were admixed (fig. S2) and subsequently discarded all SNPs that had missing data or were not polymorphic among the remaining 5 Lakota Sioux individuals. This left us with data from 2259 SNP loci to base the final analyses on.

Error estimation
We used the error estimation method from Orlando et al. (32) implemented in ANGSD (Analysis of Next Generation Sequencing Data) (33) to estimate base type–specific error rates for sequencing data from the Sitting Bull hair sample. This method relies on a comparison of the number of derived alleles in the sequencing data from the individual of interest and in a high-quality genome. The idea behind it is that any human genome should, on average, have the same number of derived alleles, and therefore, the excess of these alleles in the sequencing data from the individual of interest compared to a high-quality genome can be assumed to be due to sequencing errors and thus used as a basis for error estimation.

MDS analysis
We performed MDS analyses using the software tool bammds (version: bammds_20140602) (31). We applied the tool to the quality-filtered sequencing data from the Sitting Bull hair sample in the form of a bam file combined with the HGDP dataset that is made available along with the bammds software tool.

Admixture analyses
We used the program ADMIXTURE (34) to identify admixed individuals. When running the analyses, we assumed that the number of ancestral populations, $K$, is 2. We ran ADMIXTURE 50 times with different starting values and the difference in likelihood units between the highest likelihood and 10th highest likelihood was less than 0.000001, suggesting that convergence was achieved.

Simulation study
To be able to properly interpret the $F_{rel}$ estimate obtained for Sitting Bull and Ernie LaPointe and to estimate how powerful the new estimation method is, we simulated 100,000 datasets and applied the method to these datasets.

Data simulations
Each of the 100,000 was simulated so it mimicked the data from the analyses of the real data from of Sitting Bull and Ernie LaPointe. Specifically, each dataset consisted of low-depth sequencing data from one historical individual, SNP chip genotype data from four reference individuals, and SNP chip genotype data from two individuals for which we are interested in estimating $F_{rel}$: one, $E_{unrelated}$ that is unrelated to the historical individual and one, $E_{great-grandchild}$ that is the great-grandchild of the historical individual. Further, the number of simulated loci was chosen so the number of SNP that are polymorphic among the four reference individuals and the individual of interest (i.e., the five individuals that were used for allele frequency estimates when applying our method to the dataset) was, on average, approximately equal to the number of SNP loci used in our real data analyses, namely, 2259.

The data for each dataset were simulated by first sampling population allele frequencies from a uniform distribution between 0.1 and 0.9 (mimicking that we have data from loci on a SNP chip) for a preset number of SNP loci. Then, using these population allele frequencies, we simulated a pool of 15 haplotypes. Assuming that our loci are few and far between and thus independent, we simulated each of these haplotypes by independently for each locus sampling an allele from a Bernoulli distribution using the population allele frequency for that locus. Last, based on these haplotypes, we simulated

1) The low-depth sequencing from the historical individual by sampling two haplotypes from the pool of haplotypes without replacement and then randomly sampling an allele from one of these two haplotypes at each locus

2) The SNP chip genotype from four reference individuals by sampling four pairs of haplotypes from the pool of haplotype without replacement
3) The SNP chip data from \( F_{\text{unrelated}} \) by sampling two haplotypes from the haplotype pool without replacement

4) The SNP chip data from \( F_{\text{great-grandchild}} \) by simulating first a child of the historical individual based on the two haplotypes of the historical individual, then a grandchild based on the two haplotypes of the child, and lastly a great-grandchild based on the two haplotypes of the grandchild. More specifically, we simulated each of these offspring by sampling one new haplotype from the pool of haplotypes and combining this with a haplotype that was a recombination of the two haplotypes of the parent (e.g., the historical individual).

The recombination was simulated by randomly sampling one allele at each locus from the two haplotypes.

When simulating the datasets, we assumed that the genotypes from the present-day individuals can be obtained reliably without any errors. In contrast, we assumed a fixed error rate of 0.001 for the historical individual and added errors by simply switching the simulated sampled allele to the other allele segregating at this locus randomly with probability equal to the error rate. We used an error rate of 0.001 to mimic the error rate estimated for nontransition loci that is below 0.001 across all categories (cf. fig. S1).

**Estimating \( F_{\text{rel}} \) from the simulated datasets**

For each of the simulated datasets, our \( F_{\text{rel}} \) estimation method was applied to (i) the simulated data from the historical individual, (ii) the simulated data from the individual, \( F_{\text{unrelated}} \) or \( F_{\text{great-grandchild}} \) for which we want to estimate \( F_{\text{rel}} \), and (iii) allele frequencies estimated from four reference individuals and \( F_{\text{unrelated}} \) or \( F_{\text{great-grandchild}} \) (i.e., the same approach as we have used for Sitting Bull and Ernie LaPointe). The estimation method was implemented in R using the optimum function with the optimization algorithm set to L-BFGS-B and lower and upper bound for \( F_{\text{rel}} \) set to 1 × 10^{-12} and to 1 − (1 × 10^{-12}), respectively.

**Estimating \( F_{\text{rel}} \) from the real data**

As described above the estimation method was implemented in R using the optimum function with the optimization algorithm set to L-BFGS-B and lower and upper bound for \( F_{\text{rel}} \) set to 1 × 10^{-12} and to 1 − (1 × 10^{-12}), respectively. When estimating \( F_{\text{rel}} \) from the real data, we used genotypes from all the five unadmixed Lakota Sioux individuals for the allele frequency estimation, i.e., we included the individual for which \( F_{\text{rel}} \) was estimated in the allele frequency estimation.

**SUPPLEMENTARY MATERIALS**

Supplementary material for this article is available at https://science.org/doi/10.1126/sciadv.abh2013.

View/request a protocol for this paper from Bio-protocol.
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