Mitochondrial and Nuclear Genes Present Conflicting Portraits of Human Origins

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Human mitochondrial DNA (mtDNA) sequences reveal an abundance of polymorphic sites in which the frequencies of the segregating bases are very different. A typical polymorphism involves one base at low frequency and the other base at high frequency. In contrast, nuclear gene data sets tend to show an excess of polymorphisms in which both segregating bases are at intermediate frequencies. A new statistical test of this difference finds significant differences between mtDNA and nuclear gene data sets reported in the literature. However, differences in the polymorphism patterns could be caused by different sample origins for the different data sets. To examine the mtDNA–nuclear difference more closely, DNA sequences were generated from a portion of the X-linked pyruvate dehydrogenase Ela subunit (PDHA1) locus and from a portion of mitochondrial control region I (CRI) from each of eight individuals, four from sub-Saharan Africa. The two genes revealed a significant difference in the site frequency distribution of polymorphic sites. PDHA1 revealed an excess of intermediate-frequency polymorphisms, while CRI showed an excess of sites with the low-high frequency pattern. The discrepancy suggests that mitochondrial variation has been shaped by natural selection, and may not be ideal for some questions on human origins.

Introduction

A distribution of DNA polymorphism frequencies can be used to ask questions about recent natural selection and changing population sizes (Tajima 1989a, 1989b; Rogers and Harpending 1992; Fu and Li 1993). When a site is polymorphic, usually just two bases are observed in a sample of sequences from a population, and it is possible that one base may have a low frequency (and thus the other base a high frequency) or both bases may have intermediate frequencies. Repeatedly, studies of human mitochondrial sequences have revealed a relative abundance of polymorphisms with the low-high pattern, and a paucity of intermediate-frequency polymorphisms (Excoffier 1990; Merriwether et al. 1991; Vigilant et al. 1991; Ruvolo, Zehr, and von Dornum 1993). This departure from neutral expectations is consistent with a model of a recently growing population (Tajima 1989a; Slatkin and Hudson 1991; Rogers and Harpending 1992) and has been regarded as evidence of a recent and rapid human population expansion, perhaps within the past 200,000 years (Stringer and Andrews 1988; Stoneking 1994). Representative large and small mitochondrial studies are shown in figure 1a and b.

If the mitochondrial polymorphism pattern is primarily due to historical patterns of population size change, then similar patterns are expected of nuclear polymorphisms. However, if natural selection on mitochondria is the primary cause of the mtDNA polymorphism pattern, then other genes may not reveal the same pattern. To compare mitochondrial and nuclear genes, it was necessary to avoid those nuclear gene studies that focused on polymorphisms that were first discovered within just a small sample (Long et al. 1990; Bowcock et al. 1991). These polymorphisms necessarily tend to appear at intermediate frequencies and are biased for the purpose of site frequency analysis. A literature search returned two large nuclear gene studies in which variation was measured uniformly for all sampled individuals (fig. 1c and d). The nuclear gene studies have an excess of intermediate frequency polymorphisms while the mitochondrial studies have an excess of polymorphisms with the low-high pattern. These differences could be due to the different populations sampled for the different studies. There exists considerable population structure in global patterns of human genetic variation (Cavalli-Sforza, Menozzi, and Piazza 1993). In particular, populations from Africa are the most variable and seem to possess some of the oldest variation (Merriwether et al. 1991; Vigilant et al. 1991). About two thirds of the sampled sequences for figure 1a and b had African origins. In contrast, each of the nuclear gene studies in figure 1c and d included just one non-African population.

To address whether the differences between mitochondria and nuclear genes are due to different sample origins, a small sample of African and non-African individuals were studied at a portion of the nuclear genome and a portion of the mitochondrial genome.

Materials and Methods

DNA Sequencing

All DNA templates were from males to avoid sequencing diploid genomic templates. Sequence names and sources of genomic DNA: H-NA1, Coriell Institute for Medical Research, cell line repository # NA05963C, Amish Pedigree 884; H-NA2, Coriell Institute repository # NA07016, Utah pedigree K-1331; H-V1, Coriell Institute repository # NA11035, Venezuelan pedigree 104; H-V2, Coriell Institute repository # NA04477, Venezuelan pedigree 102; H-B1, Bantu genomic DNA male 1; H-B2, Bantu genomic DNA male 2; H-K1, Khoisan genomic DNA male 1; H-K2, Khoisan genomic
Fig. 1.—The frequency distribution of polymorphic sites. For each data set, \( n \) is the number of sequences; \( L \) is the length of the DNA sequence; and \( s \) is the number of polymorphic base pair positions, not including insertions and deletions. For comparison across data sets, \( \pi \) and \( \Theta \) have been divided by \( L \) to give estimates of diversity per base pair. Nearly all polymorphisms involve just two of the four nucleotide bases. To avoid redundancy, only the frequency of the less common base is shown. Thus, a polymorphism with one sequence that differs from the remaining \( n-1 \) sequences is counted in frequency class 1. For the few three- and four-base polymorphisms, the frequencies of the two or three least common bases, respectively, are also included in \( s \). For the large data sets, frequency classes are in groups of five. White bars are observed frequencies, and total bar height sums to \( s \). Black bars are expected levels under constant population size (Tajima 1989b). Hatched bars are expected levels found by simulations for a recently growing population (see Materials and Methods and Results). a, Mitochondrial cytochrome oxidase subunit II gene (Ruvolo, Zehr, and von Dornum 1993). b, Mitochondrial control region I and II data (Vigilant et al. 1991). c, \( \beta \)-globin data collected from the Vanuatu islands (Fullerton et al. 1994). d, Elastin data from a British Caucasian population (Raybould, Bierley, and Hulten 1995). \( L \) is not given for Elastin, which is restriction site data, and \( \pi \) and \( \Theta \) have not been standardized in this case.

DNA male 2. For the X-linked pyruvate dehydrogenase El\( \alpha \) subunit (PDHA1) locus, sequences were obtained by polymerase chain reaction (PCR) amplification using primers corresponding to base numbers 14084–14103 and 16387–16406 (Koike et al. 1990). The region sequenced corresponds to the span from position 14210 to position 15957 of the reference sequence (Koike et al. 1990). For the mitochondrial control region I (CRI), initial amplification primers corresponded to positions 15975–15996 and 16401–16422 (Anderson et al. 1981). PCR products were converted to single-stranded form (Higuchi and Ochman 1989) and sequenced directly. This was done for both strands. The GenBank accession numbers are U75933–U75940 for the PDHA1 sequences, and U75941–U75948 for the CRI sequences.

Simulations

Conventional coalescent simulations (Hudson 1990) were used to determine the distribution of Tajima’s \( D \) (Tajima 1989b), and the distribution of the difference, \( \Delta \), between \( D \) values (see Results). Each value of \( \Delta \) was determined from two independent, constant
population size coalescent simulations, one for each of two loci. For each locus, the mean of the observed values of $\pi$ and $\theta$ (see fig. 1 and Results) was used as an estimate of the population mutation parameter, $2\mu$ (see Results), that was input to the simulations.

Coalescent simulations were also conducted under a population growth scenario. As in the case of constant population size simulations, a distribution of $D$ was determined from multiple pairs of independent coalescent simulations. Each population growth simulation proceeded with a specified initial population size $N_0$, final population size $N_1$, and time of population growth, in generations. $t$. Rogers and Harpending (1992), in their table 1, provide estimates of these quantities based on a large worldwide sample of mitochondrial DNAs (Cann, Stoneking, and Wilson 1987). One of the benefits of coalescent simulation is that, apart from sample size, the population mutation parameter is the only required parameter, and it is not necessary to represent the actual population size in the simulation (Hudson 1990). This simplification, in the case of a coalescent simulation of sudden population growth, means that the actual estimates of $N_0$, $N_1$, and $t$ are scaled by the population size at the beginning of the simulations (i.e., $N_1$, since coalescent simulations proceed backward in time). With scaling, the estimates of Rogers and Harpending (1992) (rows 1 and 2 of their table 1) become $N_1 = 1$, $N_0 = 0.00594$, and $t = 0.01753$. The scaled population size terms, $N_1$ and $N_0$, can be applied to the mitochondrial simulations as well as to those for nuclear genes (assuming the population expansion did not affect the sexes differently). However, the scaled time parameter that is based on mitochondria cannot be applied directly to nuclear genes. Under the model, the time of population expansion was necessarily the same for all portions of the genome, and the scaled value of $t$ must be adjusted for the differences in effective population size that are expected for different portions of the genome. An X chromosome gene is expected to have about three times— and a nuclear gene about four times—the effective population size of mitochondria. In effect, the relative time of the expansion, if it were scaled by the effective population size for the nuclear gene regions, is more recent than for the case of mitochondria. For the simulations representing the X-linked locus (PDHA1), $t$ was set to 0.005843 (i.e., 0.01753/3). For the nuclear genes ($\beta$-globin and Elastin), $t$ was set to 0.004383 (i.e., 0.01753/4). The coalescent simulations also require a population mutation parameter. However, $\pi$ and $\theta$ (see fig. 1 and Results) are only appropriate estimates if the population size is constant, and they are not appropriate for the case of recent population expansion. To generate estimates under this model, preliminary population growth simulations were conducted (using $N_1$, $N_0$, and $t$ as described above) to find values of the population mutation parameter that generated, on average, the same number of polymorphic sites that were observed in the data sets. These values are 22 for PDHA1, 102 for the $\beta$-globin data set, 24.5 for the Elastin data set, 42 for CRI, 25 for the mitochondrial COII data set, and 170 for the large data set on mitochondrial control regions I and II. The results from these simulations were also used to generate expected site frequency distributions under the model of rapid population expansion (fig. 1).

Results

Figure 2 shows the DNA sequence polymorphisms found at PDHA1 and CRI. Polymorphism summaries and site frequency distributions are shown in figure 3. The amount of base pair heterozygosity at PDHA1 ($\pi = 0.00113$) is very close to the average of 0.0011 reported for four-fold degenerate sites within amino-acid-coding regions (Li and Sadler 1991). However this value is higher than for most X chromosomal genes, which on average appear to have about one third as much variation as autosomal genes (Hofker et al. 1986). Four polymorphisms were found at PDHA1, all within introns, and all among the four African sequences (H-B1, H-B2, H-K1, and H-K2). This pattern is roughly consistent with an out-of-Africa model of human origins. Interestingly, all four polymorphisms were also at intermediate frequency. In contrast, most of the CRI polymorphisms have the low-high frequency pattern. In short, the PDHA1 and CRI data reveal the same kind of difference (fig. 3a and b) that had been found in the other data sets (fig. 1).

A conventional chi-square test for differences in the frequency distributions cannot be made because linkage causes nonindependence among the polymorphisms within a locus. Instead, a simple two-step approach was developed: for each gene, the departure from null model expectations was assessed using Tajima’s $D$ (Tajima 1989b), and then the statistical significance of the difference between $D$ values for pairs of genes was determined. $D$ is proportional to the difference between two commonly used measures of nucleotide variation, $\pi$ and $\theta$. Under neutrality and constant population size, both measures have an expected
value that is equal to the population mutation parameter $2\theta u$ (often referred to as $\theta$), where $G$ is the effective number of gene copies in the population at the locus under investigation, and $u$ is the mutation rate at the locus per generation. $\pi$ is the average number of differences between all pairs of sequences (Nei and Tajima 1981). A polymorphism of intermediate frequency appears as a difference between many pairs of sequences, and so contributes more to $\theta$ than does a polymorphism with the low-high frequency pattern. In contrast, $\pi$ is a function only of the number of DNA sequences ($n$) and the number of polymorphic sites ($s$), and it does not depend on the frequency of polymorphisms (Hudson 1990). For each of the mtDNA data sets (figs. 1a, 1b, and 3b), the excess of low-frequency polymorphisms results in a value of $\theta$ that is less than $\theta$, and $D$ is negative. In the case of the nuclear genes (figs. 1c, 1d, and 3b) there is an excess of intermediate frequency polymorphisms compared to the expected values, and $D$ values are positive.

Let $\Delta_{n_1,n_2} = D_{n_1} - D_{n_2}$, where $n_1$ and $n_2$ are the sample sizes in the two data sets. For convenience, the $D$ values are ordered so that $\Delta_{n_1,n_2}$ is positive. If the loci are unlinked, $D_{n_1}$ and $D_{n_2}$ are independent and the probability density of $\Delta_{n_1,n_2}$ can be found using a convolution of the probability densities for the two $D$'s (expression 47 in Tajima 1989b). The general form of the convolution, and the probability that $\Delta_{n_1,n_2}$ takes a particular value $\delta$, is

$$P(\delta) = \int_{-\infty}^{\infty} \phi_{n_1}(x)\phi_{n_2}(x - \delta) \, dx,$$

where $\phi_{n_1}(x)$ is the probability that $D_{n_1}$ is equal to $x$ (eq. 47 in Tajima 1989b). The probability $p$ of observing a value that is more extreme than is observed can be found by evaluating

$$p = \int_{-\Delta_{n_1,n_2}}^{\Delta_{n_1,n_2}} P(\delta) \, d\delta + \int_{-\infty}^{-\Delta_{n_1,n_2}} P(\delta) \, d\delta.$$

By considering values greater than or equal to the observed value of $\Delta_{n_1,n_2}$ and less than or equal to $-\Delta_{n_1,n_2}$, the test becomes two-tailed. Both expressions (1) and (2) were evaluated numerically to determine the statistical significance of observed values of $\Delta_{n_1,n_2}$. Because the theoretical distribution of $D$ is actually an approximation (Tajima 1989b), a simulated distribution of $\Delta_{n_1,n_2}$ was also used. The results of the statistical tests are shown in table 1. With the exception of the contrasts involving the Elastin data set, all of the tests that include a mitochondrial gene study and a nuclear gene study are statistically significant.

These tests are based on the assumption of a constant population size (Tajima 1989b). However, given the good fit between the mitochondrial data and models of population growth (Rogers and Harpending 1992; Takahata 1993), null models of population expansion were also considered. If a population is growing very rapidly, then the gene tree histories will approach a star phylogeny (Tajima 1989a; Slatkin and Hudson 1991). Under a star phylogeny, $\theta$ and $\pi$ have a simple relationship to one another. Consider a star phylogeny connecting $n$ sequences, and let there be $x_i$ mutations on the branch leading to sequence $i$. Then the number of polymorphic sites, $s$, is equal to

$$\sum_{i=1}^{n} x_i,$$

and the average number of pairwise differences, $\pi$, is equal to
Table 1

<table>
<thead>
<tr>
<th>Mitochondrial Data Sets</th>
<th>Nuclear Gene Data Sets</th>
<th>(A) Values and Tests of Significance</th>
</tr>
</thead>
<tbody>
<tr>
<td>CR1</td>
<td>PDHA1</td>
<td>(\beta)-globin (Fullerton et al. 1994)</td>
</tr>
<tr>
<td>(\Delta)</td>
<td>2.781</td>
<td>2.866</td>
</tr>
<tr>
<td>(p)</td>
<td>0.0447*</td>
<td>0.0391*</td>
</tr>
<tr>
<td>(p)-simulated</td>
<td>0.0278*</td>
<td>0.0228*</td>
</tr>
<tr>
<td>(p)-growing population</td>
<td>0.0042**</td>
<td>0.0016**</td>
</tr>
<tr>
<td>COII (Ruvolo, Zehr, and von Dormann 1993)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>(\Delta)</td>
<td>2.803</td>
<td>2.888</td>
</tr>
<tr>
<td>(p)</td>
<td>0.0417*</td>
<td>0.0364*</td>
</tr>
<tr>
<td>(p)-simulated</td>
<td>0.0220*</td>
<td>0.0208*</td>
</tr>
<tr>
<td>(p)-growing population</td>
<td>0.0022**</td>
<td>0.0084**</td>
</tr>
<tr>
<td>Control regions I and II (Vigilant et al. 1991)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>(\Delta)</td>
<td>3.161</td>
<td>3.246</td>
</tr>
<tr>
<td>(p)</td>
<td>0.0214*</td>
<td>0.0204*</td>
</tr>
<tr>
<td>(p)-simulated</td>
<td>0.0076**</td>
<td>0.0124*</td>
</tr>
<tr>
<td>(p)-growing population</td>
<td>0.0210*</td>
<td>&lt;0.0002**</td>
</tr>
</tbody>
</table>

\(* P \leq 0.05. \\
** P \leq 0.01. \\
\(A\) is the absolute value of the difference between \(D\) for a nuclear gene data set and \(D\) for a mitochondrial gene data set. \(D\) values are given in figures 1 and 4. \\
\(p\) is the probability of observing a value greater than the observed \(A\), plus the probability of observing a value less than negative \(A\) (a two-tailed test), calculated with expression (2). \\
\(p\)-simulated is based on 5,000 independent simulated values of \(A\) assuming a constant population size (see Materials and Methods). \\
\(p\)-growing population is based on 5,000 independent values of \(A\) assuming a recent population expansion (see Materials and Methods). \\

\[
\frac{\sum_{i=1}^{n} \sum_{j=1}^{n} x_i + x_j}{\binom{n}{2}}.
\]

Substitution of (3) into (4) shows that \(\pi = 2s/n\). Furthermore, since

\[
\hat{\theta} = s \sum_{i=1}^{n-1} \frac{1}{i}
\]

(Hudson 1990), it follows that

\[
\pi = \frac{2\hat{\theta}}{n}.
\]

Thus, under a star phylogeny, \(\pi\) and \(\hat{\theta}\) are related by a scalar and will have a correlation of one. In contrast, under a constant population size, \(\pi\) and \(\hat{\theta}\) can vary with respect to each other (Tajima 1989b). This means that the quantity \(\pi - \hat{\theta}\), to which \(D\) is proportional, has a source of variation under a constant population size that is not present when a population is growing rapidly. In short, the two extremes of very rapid population growth and constant population size are associated with relatively low and high variances for \(D\), respectively. It follows that the observed values of \(D\) would be judged even more unlikely under the rapid-growth model than under the constant-population-size model.

As an additional check, simulations were conducted under a specific rapid-population-growth model that had previously been fitted to mitochondrial data. The simulations incorporated values of a recent large population size, an initial small population size, and a time of population size change that had previously been estimated (Rogers and Harpending 1992) using a large worldwide mtDNA data set (Cann, Stoneking, and Wilson 1987). The probabilities associated with the observed values of \(D\) are considerably less under this population-growth model (table 1), as expected if the variance of \(D\) is less than under a constant-population-size model.

Discussion

These analyses suggest that variation at nuclear genes and mtDNA are not both consistent with a common demographic history. In particular, the PDHA1–CRI contrast effectively eliminates simple explanations based on the genetic structure among human populations, although it remains possible that more complex demographic models could be found to fit the data.

The statistically significant findings arise even in contrasts among genes for which sample sizes and estimates of \(2\hat{u}\) were quite low. There appear to be two reasons for this. First is the simple reason that the observed values of \(A\) are large. Second is that the variance of \(D\) is essentially not a function of either sample size or \(2\hat{u}\). Indeed, for samples with more than five sequences and \(2\hat{u}\) greater than one (for the entire sequenced region, not per base pair) the mean and variance of \(D\) are nearly constant and close to zero and one, respectively (Tajima 1989b). Since \(D\) is only slightly a function of sample size and \(2\hat{u}\), the same goes for \(\Delta\). In short, statistical tests based on \(D\) and \(\Delta\) do not require large sample sizes.

In principle, natural selection at either nuclear or mitochondrial genes (or both) could be responsible for the observed differences. For example, balancing selection at nuclear genes will increase the levels of intermediate-frequency polymorphisms and elevate \(D\) for these loci. Indeed, PDHA1 does reveal more variation than most X-chromosomal genes (Hofker et al. 1986) and it is possible that this is due to balancing selection. However, the mtDNA is the most likely candidate for a selective explanation for the observed values of \(\Delta\). In contrast to the three unlinked nuclear genes, the \(D\) values from the different mitochondrial studies are not independent because the mtDNA is a single linkage group. The mtDNA is also an excellent candidate for a by-product of natural selection that occurs when multiple loci are tightly linked. Under tight linkage, natural selection cannot act effectively on multiple polymorphic
sites, and this leads to a reduction of the effective number of gene copies experienced by the region of tight linkage (Felsenstein 1974). Recent analyses of mitochondrial variation in several organisms, including humans, suggest that they are segregating an excess of slightly deleterious variation in protein-coding genes (Nachman, Boyer, and Aquadro 1994; Rand, Dorfsman, and Kann 1994; Nachman et al. 1996). These mutations are expected to create a background selection effect (Charlesworth, Morgan, and Charlesworth 1993) for the entire mitochondrial genome, simultaneously knocking down the effective population size for mitochondria and shifting the site frequency distribution toward an excess of the low–high pattern, as is observed. Background selection against strongly deleterious mutations probably could not explain the observation (Charlesworth, Morgan, and Charlesworth 1993; Charlesworth, Charlesworth, and Morgan 1995). This type of selection has only a small effect on $D$, and the mitochondrial genome has relatively few genes that present targets for strongly deleterious mutations. However, deleterious mutations of very slight effect, or that are nearly or completely recessive in their effect, can lead to a shift in the site frequency distribution and a negative $D$ (Charlesworth, Charlesworth, and Morgan 1995).

Recent data from the human Y chromosome also do not have the same pattern of polymorphism found in mitochondria. A diverse sample of 16 individuals revealed $\pi$ greater than $\theta$ (Hammer 1995). As with the other nuclear genes, $D$ is positive (0.2306) but it is not significantly different from the mitochondrial genes. The Y chromosome region that was sequenced was from the nonrecombining region, so it may also be subject to an interaction of linkage and natural selection.

The patterns of mtDNA variation have previously been interpreted as evidence of rapid population growth that followed a period of small population size about 200,000 years ago (Cann, Stoneking, and Wilson 1987; Vigilant et al. 1991), and the analysis of Rogers and Harpending (1992) suggested a date between 60,000 and 120,000 years. This historical portrait is consistent with the estimated dates of the earliest modern human fossils and has provided support for the hypothesis of a recent African origin for modern humans (Stringer and Andrews 1988). The PDHA1 polymorphisms are consistent with the African-origin portion of this hypothesis as they were found among the sequences of African origin, suggesting that African populations are more variable at this locus. However, the site frequency distributions of the nuclear genes are in conflict with the population expansion scenario. The positive $D$ values from nuclear genes suggest that human populations were relatively large and not subject to population bottlenecks during the time that human ancestors evolved into modern form.

Acknowledgments

I thank M. Bulmer, D. Rand, A. Rogers, P. Smouse, M. Stoneking and N. Takahata for helpful comments, and N. Leahy and A. Zaidi for laboratory assistance. The Bantu and Khoisan DNAs were kindly supplied by M. Stoneking and H. Soodyall. This work was supported by a Johnson & Johnson Discovery Award and by National Science Foundation grant DEB-9306625.

LITERATURE CITED


Nachman, M. W., S. N. Boyer, and C. F. Aquadro. 1994. Nonneutral evolution at the mitochondrial NADH dehydro-


Charles F. Aquadro, reviewing editor

Accepted October 31, 1996