# Why is there so little intragenic linkage disequilibrium in humans?

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#### Summary

The efficient design of association mapping studies relies on a knowledge of the rate of decay of linkage disequilibrium with distance. This rate depends on the population recombination rate, C. An estimate of C for humans is usually obtained from a comparison of physical and genetic maps, assuming an effective population size of approximately  $10^4$ . We demonstrate that under both a constant population size model and a model of long-term exponential growth, there is evidence for more recombination in polymorphism data than is expected from this estimate. An important contribution of gene conversion to meiotic recombination helps to explain our observation, but does not appear to be sufficient. The occurrence of multiple hits at CpG sites and the presence of population structure are not explanations.

## 1. Introduction

A major goal of human genetics is the identification of loci that contribute to the risk of non-Mendelian diseases. Theoretical studies have demonstrated that association mapping may be a powerful approach if the variants typed include the disease-susceptibility locus itself (Risch & Merikangas, 1996). However, short of genotyping all variable markers in the genome, association studies may not include the diseaseassociated site. The aim is then to find a single nucleotide polymorphism (SNP) in linkage disequilibrium with the susceptibility allele. The power of such association studies will depend on the regional patterns of linkage disequilibrium (Long & Langley, 1999). The efficient design of association mapping studies will therefore be aided by a knowledge of factors influencing linkage disequilibrium (cf. Kruglyak, 1999).

Linkage disequilibrium (LD) is shaped by numerous forces including genetic drift, natural selection and the population recombination rate. Consider, for example, the model commonly used to estimate population parameters: a randomly mating population at mutation–drift equilibrium (Kimura, 1968). Under this set of assumptions, and an infinite-sites mutational model, the mean squared correlation between sites is given by  $\frac{1}{(1+4N_er)}$ , where  $N_e$  is the diploid effective population size of the species and r the per generation rate of recombination between sites (Ohta & Kimura, 1971).

Thus, a key parameter in the design of association studies is the *population* recombination rate,  $C = 4N_{o}r$ (Long & Langley, 1999). C can be estimated in two ways: In the first, r is estimated from a comparison of genetic and physical maps while  $N_{o}$  is estimated from levels of diversity (e.g., Kruglyak, 1999); this method relies on the standard neutral model outlined above and an estimate of the mutation rate per generation. In what follows, we refer to this estimate of C as  $C_{max}$ . Alternatively, r and  $N_{e}$  can be estimated jointly, from nucleotide sequence polymorphism data (e.g. Hudson, 1987; Griffiths & Marjoram, 1996; Wall, 2000). In this approach, the observed patterns of linkage disequilibrium are used to estimate C. If the assumptions behind these methods are realistic, the two estimation methods should yield similar results (Hudson, 1987; Andolfatto & Przeworski, 2000).

Here, we present a comparison of the two methods of estimation of C for the nine suitable human data sets currently available. We highlight a systematic discrepancy between these two methods for both a model of constant population size and one of long-

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term exponential growth. There appears to be more evidence for recombination in polymorphism data than would be expected from  $C_{map}$ . A similar pattern has been found in other data (K. Ardlie & L. Kruglyak, personal communication).

## 2. Methods

We examine all nuclear sequence data sets for which frequency and haplotype data were available, the sample size (n) was at least 10, and the number of segregating sites (S) was at least 10. Only biallelic polymorphisms (both point mutations and indels) are included. We exclude mutations that overlap with deletions, since these offer incomplete information. For data sets other than Zfx, each individual is sequenced for the whole length (rather than at preselected variants); the data for Zfx was obtained by SSCP (for details see Przeworski et al., 2000). Haplotypes were determined by the method of Clark (1990) combined with allele-specific PCR (see Clark et al. 1998) for Ace, ApoE and Lpl. For the two other autosomal loci, they were determined experimentally ( $\beta$ -globin, Duffy gene sequence).

We use the largest world-wide sample available for each locus, even though some loci show substantial differentiation between populations. In particular, the sub-Saharan sample at *Duffy* locus is fixed for an allele that appears to be absent from other populations. Since the allele confers resistance to *Plasmodium vivax*, the African pattern of polymorphism is thought to reflect the action of natural selection (Hamblin & Di Rienzo, 2000). If we use only the Italian sample to estimate *C*,  $C_{HRM}$  increases (results not shown), so using a world-wide sample is conservative for our purposes.

Rates of recombination, r, are estimated from a comparison of the Genethon genetic map (Dib et al., 1996) and the GB4 radiation hybrid map (Gyapay et al., 1996). To estimate the regional recombination rate, we use the ratio of the genetic distance to the physical distance between the two (high confidence) microsatellite markers closest to the locus of interest. This method gives us an estimate of the number of cM/cR. To obtain an estimate of the number of cM/Mb, we use chromosome-specific conversion factors for cR to Mb (Hudson et al., 1995). Recombination rates are often estimated using more than two microsatellite markers to reduce the sampling variance associated with a small number of meioses. However, if rates vary on the scale of a couple of megabases, rates averaged over longer distances may not be informative locally. In addition, random error in r estimation will not lead to a systematic trend. Thus, for the purposes of this study, we choose to use only two markers. Estimates using other methods are listed in Table 1.

To obtain an estimate  $C_{map}$  of the population recombination rate *C*, we multiply *r* by  $4N_e$  (or  $2N_e$  if X-linked).  $N_e$  is assumed to be  $10^4$  (Li & Sadler, 1991; Takahata, 1993). This estimate of  $N_e$  is obtained from observed levels of diversity; it relies on assumptions of a panmictic population of constant size with no selection, and requires an estimate of the mutation rate per generation (see Section 3 for more details).

Estimates of C based on polymorphism data have high variances and are often biased (Wall, 2000). We chose an estimate (referred to as  $C_{HRM}$ ) that appears to be roughly unbiased under the standard neutral model outlined above, has relatively low mean squared error, and can be calculated for large polymorphism data sets (Wall, 2000; J.D.W., unpublished results).  $C_{HRM}$  is a maximum likelihood estimator of C based on two summaries of the data:  $R_M$ , the minimum number of recombination events (cf. Hudson & Kaplan, 1985), and H, the observed number of distinct haplotypes (for more detail see Wall, 2000). The values of H may be underestimates for Lpl and Ace, due to incomplete phase information. For Ace, the presence of singletons on individuals C08, C09 and C24 (Rieder et al., 1999) increased the minimum number of inferred haplotypes to H = 16. The likelihood L (C| $R_M$ , H) is estimated from coalescent simulations (Kingman, 1982; Hudson, 1990). At least  $2 \times 10^5$  replicates are run for each parameter combination. To determine credibility intervals for the maximum likelihood estimates, we employ the standard  $\chi^2$  approximation for the likelihood ratio statistic  $-2\ln(L_1/L_0)$ , where  $L_0$  is the maximum likelihood and  $L_1$  is the likelihood at an alternative point. We caution that there is no direct evidence that the standard  $\chi^2$  approximation is appropriate; future work will address this issue. Loci are assumed to be independent, so the likelihood of the whole data equals the product of the likelihoods at each locus. If the estimated likelihood for a particular locus is 0, we replace it with the reciprocal of the number of trials run to keep all calculations well defined. Most simulations were run using modifications of programs kindly provided by R. Hudson.

We assume a neutral infinite-sites model for our simulations. Our approach differs slightly from standard coalescent simulations: instead of conditioning on a population mutation rate, we generate genealogies and place the observed number of mutations, S, on the tree (cf. Hudson, 1993). Standard full likelihood methods exist that use all the information in the data (e.g., Griffiths & Marjoram, 1996), but they are computationally prohibitive.

Besides the standard, equilibrium panmictic null model, we consider two alternative demographic models. To incorporate recent population growth, we consider a model with a constant population size followed by exponential growth (cf. Marjoram &

Locus	$n^a$	Bps	$S^b$	$C_{_{HRM}}$	$C_{map}$	Other $C_{map}$ estimates	Obs. $R_{M}$	$P^c$	References
β-globin	349	2670	19	24.0	2.4	$2 \cdot 0^d, 3 \cdot 1^e$	3	0.039	Harding et al. (1997)
Duffy	82	1931	16	5.0	0.5	$1 \cdot 0^d$	2	0.010	Hamblin & Di Rienzo (2000)
Lpl	142	9700	87	120.0	9.3	$4 \cdot 2^{f}, 4 \cdot 5^{g}, 8 \cdot 8^{d}$	22	$< 10^{-5}$	Clark et al. (1998)
Åce	22	24000	78	17.0	9.1	$1.6^{e}, 7.6^{d}$	6	0.153	Rieder et al. (1999)
ApoE	192	5491	22	23.0	6.5	$3\cdot 3^f$	8	$< 10^{-5}$	Fullerton et al. (2000)
Dmd44	41	3000	17	60.0	2.1	$0.1^{e}, 0.8^{g}, 2^{d}, 3.6^{h}, 4.6^{i}$	7	$< 10^{-5}$	Nachman & Crowell (2000 <i>a</i> )
Pdha1	35	4200	24	6.2	5.0	$0.2^{e}, 0.8^{g}, 3.2^{i}, 3.7^{d}$	3	0.148	Harris & Hey (1999)
$Xq13\cdot 3$	70	10163	33	2.0	0.3	$0.3^{j}, 1.7^{d}$	1	0.239	Kaessmann et al. (1999)
Źfx	336	1089	10	3.8	0.1	$0.04^e, 0.7^g, 0.8^d$	1	0.030	Jaruzelska et al. (1999)

Table 1. Conflicting estimates of the population recombination rate in humans

<sup>*a*</sup> Number of chromosomes sequenced.

<sup>b</sup> Number of segregating sites.

<sup>c</sup>  $P = \Pr(R_M \ge obs. R_M | C_{map}).$ 

<sup>*d*</sup> B. Payseur, personal communication. Rates are obtained from a comparison of the Genethon genetic map with the GB4 radiation hybrid map, using a sliding window approach of 5 microsatellites on each side of the locus of interest and cR-Mb chromosome-specific conversion factors from Hudson *et al.* (1995). A linear function was fit to estimate recombination rates. For details, see Payseur and Nachman (2000).

<sup>e</sup> Huttley *et al.* (1999).

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<sup>f</sup> S. M. Fullerton, personal communication. Per band estimates of recombination rate were obtained from Morton's integrated database (Collins *et al.* 1996), averaged across threeband windows.

<sup>g</sup> B. Payseur, personal communication. Rates are estimated from the Morton integrated maps as in <sup>c</sup>.

<sup>*h*</sup> Nachman & Crowell (2000*a*).

<sup>*i*</sup> Nachman *et al.* (1998).

<sup>j</sup> Kaessman *et al.* (1999).

Table 2. Est	imates of th	e population	recombination	rate under	a model o	f exponential	growth
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Locus	$C_{map}{}^a$	Constant size $C_{_{HRM}}$	Exp. growth <sup>b</sup> $C_{_{HRM}}$	Exp. growth <sup>e</sup> $C_{_{HRM}}$	$P^d$
β-globin	2.4	24.0	42.0	24.0	0.051
Duffy	0.5	5.0	1.7	1.9	0.016
Lpl	9.3	120.0	e	e	$< 10^{-6}$
Åce	9.1	17.0	6.3	8.4	0.291
ApoE	6.5	23.0	e	7.0	0.00003
Dmd44	2.1	60.0	35.0	27.0	$< 10^{-6}$
Pdha1	5.0	6.2	2.1	2.7	0.237
$Xq13\cdot 3$	0.3	2.0	0.5	0.7	0.334
Zfx	0.1	3.8	1.2	1.6	0.030

<sup>*a*</sup>  $C_{map}$  was estimated under the standard neutral model, so is a slight overestimate of  $4N_0r$ . <sup>*b*</sup> The onset of growth is 50 Kya, and the population size increases from  $10^4$  to  $10^6$ .

The onset of growth is 50 Kya, and the population size increases from  $N_0 = 10^4$  to  $N_1 = 10^5$ .

The  $C_{HBM}$  values reported are estimates of  $4N_0r$ .

<sup>*d*</sup>  $P = \Pr^{m,m}(R_M \ge \text{obs.} R_M | C_{map})$ . The model of growth is as in <sup>c</sup>.

<sup>e</sup> Cannot be calculated because the estimated likelihood of the data is 0 (see text).

Donnelly, 1994). The population size increases from  $N_0 = 10^4$  to  $N_1 = 10^5$  or  $10^6$ , where  $N_0$  is the effective population size 50 thousand years ago (50 Kya), and  $N_1$  is the current effective population size. The  $C_{HRM}$ values reported in Table 2 are estimates of  $4N_0r$ . To model population differentiation between African and non-African samples, we also run coalescent simulations of a symmetric two-island model (cf. Wright, 1931). The number of individuals drawn from each island corresponded to the actual sampling scheme for each locus. Migration rates were chosen to yield mean  $F_{st}$  values (Wright, 1951) of 0.15 and 0.30. Higher values of  $F_{st}$  correspond to more population differentiation. Most observed F<sub>sT</sub> values are lower than 0.15 (see, e.g., Cavalli-Sforza et al., 1994). Other aspects of the simulations were as above.

To test for the possible effects of multiple mutations at CpG sites, we rerun our likelihood simulations excluding all segregating sites where either allele forms a cg with an adjacent nucleotide site. This analysis is conservative, and excludes all transitions and transversions both to and away from CpG sites.

Finally, we consider an alternative model of recombination that incorporates both crossing-over and gene conversion. We generalize the standard coalescent with recombination (Hudson, 1990) by assuming gene conversion events (with geometrically distributed tract lengths) occur at constant rate on all the branches (cf. Wiuf & Hein, 2000). Since very little is known about gene conversion in mammals, we consider a model that is plausible based on research in Drosophila and yeast (e.g. Orr-Weaver & Szostak, 1985; Petes et al., 1991; Hilliker et al., 1994). We consider a mean tract length of 500 base pairs, and take the rate at which conversion events originate at a given base pair to be equal to the rate at which crossing over events occur at that base pair. In the

text,  $C_{HRM}$  refers to the estimated population crossingover rate. A C program written by the authors that incorporates gene conversion is available on request.

## 3. Results and discussion

As can be seen in Table 1,  $C_{HRM}$  is greater than  $C_{map}$ for 9 of 9 loci, for many loci by an order of magnitude. If either ordering of estimates were equally likely, the probability that all 9 have the same one by chance is P = 0.0039 (two-tailed). In other words, the patterns of LD suggest more recombination than does an integration of genetic and physical maps. As an illustration, when 10<sup>5</sup> coalescent simulations are run for Lpl with n = 142, S = 87 and  $C = C_{map}$ , the simulated  $R_M$  value is *always* less than the actual  $R_M$ . Since the expected  $R_M$  increases with increasing recombination rate, it is likely that  $C \gg C_{map}$  for Lpl. There is a similar situation for ApoE and Dmd44, where once again all the simulated  $R_M$  values are less than the actual value. All but one of the estimated rates  $C_{HRM}$  are greater than the approximate average for the human genome of 1 cM/Mb (e.g. Bouffard et al., 1997; Nagaraja et al., 1997). When 95% credibility intervals are constructed for  $C_{HRM}$  at each locus, the intervals exclude  $C_{map}$  for 6 of 9 loci ( $\beta$ -globin, Duffy, Lpl, ApoE, Dmd44 and Zfx). When the loci are considered together, the  $C_{map}$  values as a whole can be strongly rejected (X = 71.7;  $\chi^2$ , 9 d.f.;  $P < 10^{-11}$ ). Our finding of low levels of intragenic LD is particularly surprising in light of the high levels of pairwise linkage disequilibria that have been detected over hundreds of kilobases in some populations (e.g. Taillon-Miller et al., 2000; Wilson & Goldstein, 2000). We now discuss possible explanations for the systematic differences between  $C_{HRM}$  and  $C_{map}$ .

Table 3.	The effect	of	'a symmetric	island	l model	l on th	he minimum	number	• of	recom	bination	events,	$R_{M}$
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Locus	$n_{Africans}{}^{a}$	$n_{Non-Africans}^{b}$	$\mathrm{F_{st}}_{P^c}=0.15$ $P^c$	$\begin{array}{l} \mathbf{F}_{\mathrm{ST}} = 0.30\\ P \end{array}$
B-globin	103	246	0.033	0.025
Duffy	48	34	0.009	0.007
Lpl	48	94	$< 10^{-5}$	$< 10^{-5}$
Ace	10	12	0.116	0.072
ApoE	48	144	$< 10^{-5}$	$< 10^{-5}$
Dmd44	10	31	$< 10^{-5}$	$< 10^{-5}$
Pdha1	16	19	0.126	0.091
$Xq13\cdot3$	23	47	0.220	0.190
Zfx	114	222	0.029	0.025

 $n_{Africans}$  is the number of African chromosomes sampled in each survey.

 ${}^{h} n_{Non-Africans}$  is the number of non-African chromosomes sampled in each survey.  ${}^{e} P = \Pr(R_{M} \ge \text{obs. } R_{M} | C_{map})$ . To estimate P, we ran 10<sup>5</sup> simulations of a symmetric island model, with migrations rates chosen to yield an average  $F_{sT}$  value of 0.15 and 0.30 (cf. Takahata, 1983). P is the proportion of runs where  $R_{M}$  was greater than or equal to the observed value (listed in Table 1). In all cases, P decreases with decreasing migration rates, i.e. for a higher F<sub>ST</sub> value.

#### (i) Departures from demographic assumptions

The discrepancy between estimates of C could be due to a number of factors, including error in our estimates of  $N_{e}$  or r, variation in recombination or mutation rates, multiple mutations at the same nucleotide site, and gene conversion. In addition, the demographic model underlying the calculation of  $C_{HRM}$  may be wrong in ways that lead to a substantial overestimate of C. Since the human population is now over 6 billion, an obvious candidate is a change in effective population size over time.

Evidence for an old onset to population growth (i.e. 50-100 Kya) is equivocal, with mtDNA, Y chromosome and most microsatellite data supporting it (e.g. Rogers & Harpending, 1992; Reich & Goldstein, 1998; Kimmel et al., 1998; Gonser et al., 2000; Thomson et al., 2000) while many nuclear sequence studies do not (Hey, 1997; Harding et al., 1997; Przeworski et al., 2000; Wall & Przeworski, 2000). Table 2 presents estimates of C for a model of constant population size  $(N_e = 10^4)$  followed by 10fold growth over 50 Kya. The discrepancy between  $C_{HRM}$  and  $C_{map}$  values is smaller, and 95% credibility intervals exclude  $C_{map}$  only for  $\beta$ -globin and Dmd44. For Lpl,  $C_{HRM}$  cannot be calculated because the estimated likelihood of the data is 0 under this model of exponential growth. In other words, for any recombination rate, the observed data is extremely likely. This may be due to an underestimate of the number of distinct haplotypes, as the phase of singletons and doubletons was not established. However, it is probably due at least in part to the inherent unlikeliness of the model of growth considered. In fact, at all 9 loci the likelihoods of  $C_{HRM}$  under recent population growth are less than the corresponding likelihoods under a constant population size model.

By multiplying likelihoods across loci, we find  $\Pi L$  $(C_{HRM})$  is more than  $2 \times 10^9$  times larger under a constant size model than under our model of recent growth. If we were to fix the start time of exponential growth and consider the growth rate as a freely varying parameter, then 10-fold (or more) growth can be rejected at the 5% level for 5 of 9 loci. In the fourth column of Table 2, we list results for 100-fold growth instead of 10-fold; the qualitative conclusions are the same. In summary, even though recent exponential growth generally lowers the point estimates of  $C_{HBM}$ , the model of growth itself is extremely unlikely. Other aspects of human nuclear sequence data (e.g. the frequency spectrum of segregating mutations) also suggest that a simple model of growth is not appropriate (Wall & Przeworski, 2000).

A realistic demographic model is likely to be much more complex than models considered here (Wall & Przeworski, 2000). For example, population structure is well documented in humans (Cavalli-Sforza et al., 1994). It should be noted, however, that population structure makes it even less likely to observe unusually high values of  $C_{HRM}$  relative to  $C_{map}$ . Subdivision decreases the effective rate of recombination since haplotypes in different subpopulations will not have a chance to recombine as often as under panmixia. This was verified by simulations of a symmetric finiteisland model under a range of parameter combinations for the estimators of Hudson (1987), Griffiths & Marjoram (1996), Hey & Wakeley (1997) and Wall (2000) (results not shown). As an illustration, we run simulations under a two-island model (meant to correspond to African and non-African populations) for the 9 data sets in Table 1, with migration rates chosen to yield average  $F_{st}$  values of 0.15 or 0.30 (cf. Takahata, 1983). Table 3 shows that in all cases, the probability of observing the actual value of  $R_M$  or

greater given  $C_{map}$  decreases, making the discrepancy between  $C_{HRM}$  and  $C_{map}$  more difficult to explain.

# (ii) Errors in parameter estimation

A second explanation for the discrepancy between  $C_{HRM}$  and  $C_{map}$  is that  $N_{e}$  is underestimated. Estimates of  $N_{\circ}$  assume the standard neutral model, and are based on observed levels of diversity and an estimate of the per generation mutation rate (u). The rate u can be estimated directly, from the observed rate of spontaneous mutations, or indirectly, from divergence data, assuming a time (in generations) to the common ancestor of humans and a closely related species. The two methods yield similar estimates: when the first approach is applied to haemophilia B,  $\hat{u} = 2.14 \times 10^{-8}$ (Giannelli et al., 1999). Based on divergence between humans and chimpanzees at 18 pseudogenes,  $\hat{u} =$  $2.5 \times 10^{-8}$  (Nachman & Crowell, 2000b). The nucleotide diversity at 4-fold degenerate sites, which are thought to be evolving neutrally, is 0.11% (Li & Sadler, 1991; Cargill et al., 1999). Equating this level of diversity with  $4N_{e}u$  for the two estimates of u, we obtain  $N_{e}$  estimates of  $1.1 \times 10^{4}$  and  $1.31 \times 10^{4}$ . If these values are used instead of  $10^4$ ,  $C_{map}$  only increases by 10–31 %. To assess whether errors in estimating  $\mu$  or  $N_{\rm e}$  could explain our observation, we redo our analyses with double the value of  $C_{map}$ . Seven of 9 loci have  $C_{HRM} > 2C_{map}$ , and  $2C_{map}$  is excluded from the 95% credibility interval for 4 of 9 loci. As a whole, taking  $2C_{map}$  as the recombination parameter at each locus can still be strongly rejected (X = 50.9;  $\chi^2$ , 9 d.f.;  $P < 10^{-7}$ ).

A third possibility is that our estimates of the recombination rate per meiosis are in error. In this analysis, r is interpolated from the scale of megabases to one of kilobases. Different assumptions about the extent and scale on which rates vary across the genome lead to different choices for estimates of r. For example, if rates do not vary much and if genetic maps have large sampling errors, local estimates of r might best be obtained by averaging across many markers. In contrast, if they vary greatly, averaging over large distances might be uninformative locally. In Table 1, we list alternative estimates of r available for the loci considered; although far from exhaustive, these were obtained using a variety of maps and methods. Different methods yield rates that vary by as much as an order of magnitude. However, even if we take the highest estimate for each locus,  $C_{HRM}$  is greater than  $C_{max}$  at all 9 loci. This suggests that the pattern we observe is not the result of a particular choice of restimate.

This said, the estimates listed in Table 1 are averages over large distances. Finer-scale mapping has yielded higher estimates for both the  $\beta$ -globin 5' region and Dmd44 (see below). Whatever method is used,

estimates of r can only be extrapolated to different scales if variation in rates across the genome is negligible. Yet there is evidence of substantial variation in recombination rates across the genome at all scales (e.g. Fullerton et al., 1994; Dunham et al., 1999; Lien et al., 2000). In yeast and in maize, it has been proposed that some hotspots for recombination may be associated with promoter regions (cf. Atcheson & Easton Esposito, 1993; Lichten & Goldman, 1995; Dooner & Martinez-Ferez, 1997; Nicolas, 1998). If so, transcribed regions may experience higher rates of recombination than suggested by larger scale averages. (In Table 1, all but  $Xq13\cdot3$  are transcribed.) Interestingly,  $\beta$ -globin, Dmd44 and, more tentatively, Lpl are thought to contain a recombination hotspot (Oudet et al., 1992; Fullerton et al., 1994; Templeton *et al.*, 2000), although only for  $\beta$ -globin and Dmd44 is there evidence independent of polymorphism data. Our pattern remains even if we exclude  $\beta$ -globin and Dmd44 from our analyses. For over half of the remaining loci,  $C_{map}$  falls outside the 95% credibility interval. When the information from the 7 remaining loci is combined, the  $C_{map}$  values can be excluded with high confidence (X = 42.2;  $\chi^2$ , 7 d.f.;  $P < 10^{-6}$ ). This is still the case if all of the  $C_{map}$  values are doubled (X = 26.8;  $\chi^2$ , 7 d.f.;  $P < 5 \times 10^{-4}$ ). Thus, our observation appears to be general, rather than the result of one or two fluke hotspots.

It is also interesting to note that the opposite pattern is found in *Drosophila*, where a sequencebased estimate of C (Hudson, 1987) is systematically *below* the laboratory-based estimate of the population rate of crossing-over per physical distance (Andolfatto & Przeworski, 2000). Thus, if genes are hotspots for recombination, they may be so in some species but not others.

### (iii) Other departures from model assumptions

The simulations assume that all mutations occur at a previously unmutated site while in actual data there may be multiple hits to the same site. Violations of the infinite-sites model might lead to spurious inferences of recombination (cf. Templeton et al., 2000). For example, transitions away from CpG base pairs are thought to be 10-13 times more likely than other nucleotide base substitutions (Anagnostopoulos et al., 1999; Giannelli et al., 1999; Nachman & Crowell, 2000b; Templeton et al., 2000), and transversions away from CpG sites probably have elevated rates as well (Nachman & Crowell, 2000b). The fraction of polymorphisms that occur towards or away from CpG sites varies greatly across samples, from 0 at Zfxto 0.46 in Pdha1. What matters regarding inferences about recombination is not the number of such sites per se, but the expected number of multiple hits. Under the standard neutral model, this number can be approximated as follows: Consider Lpl, with a heterozygosity level of 0.166%, excluding indels. (This estimate is conservative for our purposes, since we do not exclude CpG sites.) We find 20 polymorphisms at CpG sites (Templeton et al. (2000) report 19). If both transitions and transversions occur at 13 times the standard rate, an estimate of the population mutation rate,  $\theta$ , away from CpG sites is roughly 2% per site (i.e.  $0.02 \approx 1.66 \times 10^{-3} \times 13$ ). If we (conservatively) assume that all mutations occur away from CpGs, then the probability of a multiple hit is  $Pr(k \ge 2 | k \ge 1)$ ,  $\theta = 0.02 \approx 0.056$  per site. Thus, the expected number of multiple hits is roughly 1. Even if the multiple hit resulted in the spurious inference of two recombination events, the true  $R_M$  would be at least 20; the  $\Pr(R_M \ge 20 | C = C_{map}) < 10^{-6}$ . Thus, the expected number of multiple hits in these data sets seems sufficiently small to be of minor concern, unless there is a large degree of variability in mutation rates between different CpG sites. Even if we are extremely conservative and throw out all mutations at CpG sites, there is no noticeable trend on  $C_{HRM}$  values: four of them decrease, three of them increase, and the remaining two stay the same. Excluding all polymorphisms that occur at CpG sites, the 95% credibility intervals excluded  $C_{map}$  for 5 of 9 loci; they are generally broader since smaller data sets contain less information.

Another important factor in shaping intralocus patterns of linkage disequilibrium is gene conversion. Large-scale estimates of the recombination rate r do not include the effects of gene conversion; yet, on the scale of a gene, there may be an important contribution of gene conversion to the overall rate of genetic exchange (Andolfatto & Nordborg, 1998; Langley et al., 2000). We test this by implementing coalescent simulations with gene conversion (see Section 2). Table 4 lists the new estimates of  $C_{HRM}$  for the specific model considered. It is not surprising that with a model with more recombination (in the form of gene conversion), the differences between  $C_{map}$  and  $C_{HRM}$ decrease. Unlike the exponential growth simulations described earlier,  $L(C_{HRM})$  generally increases when gene conversion is included;  $\Pi L(C_{HRM})$  increases 44fold with the inclusion of gene conversion. In other words, the data (independent of  $C_{map}$ ) indicate that a model of crossing-over and gene conversion actually fits the data better (i.e. is more likely) than a model of only crossing over. Both crossing-over and gene conversion tend to increase H and  $R_M$ ; however, gene conversion generally increases  $R_M$  much more than H (results not shown). In human data we observe relatively large  $R_M$  values and relatively small H values, which is support for a model of gene conversion. However, our model of gene conversion is not a sufficient explanation for our data: the 95% credibility intervals still do not include  $C_{max}$  for 5 of

Table 4. Estimates of the population recombinationrate under a model of crossing-over and geneconversion

Locus	$C_{_{HRM}}$ (with gene conversion) <sup><i>a</i></sup>	$P^{a,b}$
β-globin	12.0	0.217
Duffy	2.3	0.061
Lpl	70.0	$< 10^{-5}$
Åce	9.2	0.519
ApoE	12.0	0.001
Dmd44	31.0	$< 10^{-5}$
Pdha1	2.9	0.490
$Xq13\cdot 3$	1.2	0.312
Zfx	1.6	0.062

<sup>*a*</sup>  $C_{HRM}$  and  $C_{map}$  refer to the population crossing-over rates (see text).

<sup>b</sup>  $P = \Pr(R_M \ge \text{obs. } R_M | C_{map}).$ 

9 loci (*Lpl*,  $\beta$ -globin, *ApoE*, *Dmd44* and *Zfx*). The  $C_{map}$  values taken together can still be rejected (X = 57.0;  $\chi^2$ , 9 d.f.;  $P < 10^{-8}$ ), even if *Dmd44* and  $\beta$ -globin are excluded (X = 29.1;  $\chi^2$ , 7 d.f.;  $P < 2 \times 10^{-4}$ ). This is the case even if the ratio of conversion to crossover events is doubled (results not shown). We caution that very little is known about gene conversion in mammals, so it is unclear whether another model (e.g. one with a higher relative rate of gene conversion events), with the addition of hotspots at one or more loci, would be a sufficient explanation of our pattern.

Estimates of C will only be useful if the model on which they depend is adequate. Yet, intragenic patterns of linkage disequilibrium appear to be inconsistent with both a constant population size model and a model of long-term exponential growth (given observed levels of diversity and divergence). A more appropriate demographic model should lead to some agreement between these and pedigree-based estimates of C. In principle, patterns of LD could be used to infer what a better model of human demography might be. However, this inference relies on a knowledge of local rates of recombination. If gene conversion and recombination hotspots are frequent within transcribed regions, pedigree-based estimates of the rate of crossing-over r may not be very informative locally, and intragenic patterns of LD may not help one infer the correct demographic model. Conversely, if independent evidence (such as the frequency spectrum of segregating mutations) suggests a model for the evolutionary past of humans, LD-based estimates of C will be quite informative about the local recombinational environment.

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