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Authors
Hinch, Anjali G  
Tandon, Arti  
Patterson, Nick  
et al

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The landscape of recombination in African Americans

A full list of authors and affiliations appears at the end of the article.

Abstract

Recombination, together with mutation, is the ultimate source of genetic variation in populations. We leverage the recent mixture of people of African and European ancestry in the Americas to build a genetic map measuring the probability of crossing-over at each position in the genome, based on about 2.1 million crossovers in 30,000 unrelated African Americans. At intervals of more than three megabases it is nearly identical to a map built in Europeans. At finer scales it differs significantly, and we identify about 2,500 recombination hotspots that are active in people of West African ancestry but nearly inactive in Europeans. The probability of a crossover at these hotspots is almost fully controlled by the alleles an individual carries at \textit{PRDM9} \((P<10^{-245})\). We identify a 17 base pair DNA sequence motif that is enriched in these hotspots, and is an excellent match to the predicted binding target of African-enriched alleles of \textit{PRDM9}.

In humans and many other species, recombination is not evenly distributed across the genome, but instead occurs in “hotspots”: two kilobase (kb) segments where the crossover rate is far higher than in the flanking DNA sequence\(^1\),\(^2\),\(^3\). The highest resolution genetic map in contemporary humans to date, the “deCODE Map”, is based on about 500,000 crossovers identified in 15,000 Icelandic meioses\(^4\). However, a limitation of maps built in people of European descent\(^4\),\(^5\),\(^6\) is that they may not apply equally well in other populations, as suggested by comparisons of maps across ethnic groups\(^4\),\(^7\),\(^8\),\(^9\) and patterns of linkage disequilibrium (LD) breakdown which suggest that more of the genome may be recombinationally active in West Africans\(^10\). It is known that a major determinant of the positions of recombination hotspots is \textit{PRDM9}, a meiosis-specific histone H3 methyltransferase whose zinc finger (ZF) domain binds DNA sequence motifs\(^11\),\(^12\),\(^13\). In Europeans, \textit{PRDM9} ZF arrays are predominantly of two similar types, “A” and “B”, both of which bind the 13-bp motif CCNCCNTNNCCNC\(^11\). In contrast, 36% of West African...
alleles are not of the A or B type\textsuperscript{9,13}. Sperm typing of males who carry neither the A nor the B allele has shown no evidence of crossover activity at recombination hotspots associated with the 13-bp motif\textsuperscript{9}.

To investigate differences in the crossover landscape across human populations, we built a genetic map in African Americans, who have an average of about 80\% West African and 20\% European ancestry, leading to genomes comprised of multi-megabase stretches of either West African or European ancestry\textsuperscript{14}. Computational approaches, including HAPMIX\textsuperscript{15}, have been developed to infer the probability of 0, 1 or 2 European or African alleles at each locus in individuals genotyped at hundreds of thousands of single nucleotide polymorphisms (SNPs)\textsuperscript{15,16,17}. Positions where the inferred number of European or African alleles changes reflect crossover events that have occurred since admixture began (on average six generations ago\textsuperscript{15}). The change in the probability of European ancestry between adjacent SNPs can be interpreted as the probability of such a crossover between them. We inferred crossover events in 29,589 apparently unrelated African Americans who had been genotyped on SNP arrays in genetic association studies (Methods; Figure 1A). To minimize false-positive crossovers, we restricted to crossovers that HAPMIX inferred with probability of >95\%, and that were flanked by a minimum of 2 centimorgan (cM) stretches where the ancestry was inferred to be unchanging (Note S1). This produced 2,113,293 high-confidence crossovers, with a typical switch point resolved within 70kb with probability 50\% (Note S1).

To build a high resolution African American genetic map (AA Map), we leveraged the fact that most crossovers occur in hotspots shared across individuals\textsuperscript{1} (Methods). Intuitively, while any crossover can only be roughly localized, inter-SNP intervals that are inferred to have an appreciable probability of crossover in multiple individuals are likely to contain recombination hotspots, allowing much better localization (Figure S1). To implement this idea, we modeled the recombination rate for each inter-SNP interval as shared across individuals, and used a Markov Chain Monte Carlo (MCMC) to sample rates consistent with the data (Methods). This provides well-calibrated estimates of the crossing-over rate between all pairs of markers as well as estimates of rate uncertainty (Note S1 and Figure S2). We find that the interval size at which the average recombination rate is equal to the standard error is 6 kb, which is the same accuracy that would be expected from a map based on 500,000 crossovers whose boundaries were precisely resolved (Note S1). Despite this high resolution, there are also some limitations. First, the AA Map does not separately infer male and female recombination rates (it is a sex-averaged map) and requires normalization by the total map length (like LD maps\textsuperscript{3,18}). Second, the map has less resolution and may miss a higher fraction of true crossovers at loci where it is more difficult to detect and resolve crossovers due to low SNP density or low differentiation between West Africans and Europeans. Third, the map may be biased where ancestry deviates from the average, for example at 8q24, where the 10\% of the people in this study who have prostate cancer have an elevated proportion of African ancestry\textsuperscript{19}. Fourth, the map assumes that all individuals are unrelated, whereas in fact there is likely some shared ancestry, resulting in multiple counting of some crossovers and an overestimate of map precision.

To assess the accuracy of the AA Map, we generated an independent African American pedigree map by analyzing 222 nuclear families that included 1,056 meioses in which we
could directly detect crossovers between parent and child (Methods; Figure 1A).
Examination of the AA Map rate around directly detected crossovers confirms the high resolution: the rate around such crossovers shows at least as strong a peak as that observed in maps based on LD\(^2,3,18\) (Figure S3). We next computed correlation coefficients for both the AA Map and the deCODE Map\(^4\) to maps derived from the breakdown of LD in Europeans (CEU) and West Africans (YRI)\(^18\). At broad scales (>3 Mb) they are almost identical (\(\rho > 0.97\); Table 1) At fine scales, the AA Map is more accurate (Table 1 and Table S1), as reflected in a modest improvement in correlation to the CEU Map at a 3kb scale (\(\rho_{AA,CEU(3kb)} = 0.66\) vs. \(\rho_{deCODE,CEU(3kb)} = 0.58\)), and a major improvement for the YRI Map (\(\rho_{AA,YRI(3kb)} = 0.71\) vs. \(\rho_{deCODE,YRI(3kb)} = 0.53\)). The deCODE Map is more correlated to the CEU Map than to the YRI Map at scales <1 Mb, suggesting that this map, built in Icelanders, reflects more European recombination rates. The AA Map shows the opposite pattern, suggesting that it reflects more West African recombination patterns.

We compared the rate estimates for all four maps (AA, deCODE, CEU and YRI) over a 200 kb region within the MHC locus where recombination rates in European males have been characterized through sperm typing\(^1\) (Figure 1B). The AA Map detects five of six known hotspots, and localizes them to within 1 kb (the sixth hotspot is weak, with a peak male rate below the genome average\(^1\)). Strikingly, the two maps based on samples with African ancestry (AA and YRI) found a hotspot not present in either map based on samples of European ancestry (deCODE and CEU) (Figure 1C; Figure S4 gives a second example). We confirmed that such “African-enriched” hotspots also occur genome-wide, by examining 2,375 loci with recombination rate peaks in the YRI Map (>5 cM/Mb) but not the CEU Map (<1 cM/Mb), and finding a rate rise in the independently generated AA Map, but not in the deCODE Map (Figure S5A). In the reciprocal experiment searching for European-specific hotspots, we find no such evidence for genuine ancestry specificity; at loci with recombination rate peaks in the CEU Map but not the YRI Map, there are weak peaks in both the deCODE and AA maps (Methods; Figure S5B). Thus, hotspots active in Europeans are consistently “shared” with YRI and African Americans, while populations with African ancestry harbor additional, non-shared hotspots we call “African-enriched”.

To understand the features of recombination in West Africans that differ from Europeans, we estimated the degree to which each African American person’s crossovers occur in “African-enriched” hotspots, compared with “Shared” hotspots, a phenotype we refer to as their “African-enrichment” (AE). We view each individual’s crossovers as sampled from a mixture of two genetic maps—an “S Map” of shared hotspots based on the deCODE Map, and an “AE Map” of African-enriched hotspots that is learned from comparing the deCODE and AA Maps—so that the proportion of crossovers assigned to the AE Map is a person’s AE phenotype (Note S4). We tested approximately 3 million SNPs (genotyped and imputed) for association with three phenotypes: AE, usage of LD-based hotspots known to be enriched for the 13-bp motif CCNCCNNTNCCNC\(^20\), and genome-wide crossover rate (in pedigrees) (Methods and Note S4). In crossovers detected in unrelated African Americans, the alleles a person carries are only sometimes descended from the ancestor in whom the crossover occurred, thus adding noise to the association signal (nevertheless there is useful signal given the large sample size; Note S4). In the Pedigree Map, association between alleles and AE can be tested directly because we have genotypes in the parents.
The SNP showing the strongest association with AE is rs6889665 (P=1.5×10^{-246}; Figure 2A, Figure S6), which has a derived allele frequency of 29% in YRI and 2% in CEU, and is within 4 kb of the ZF array of PRDM9^4,9,11,12,13. This SNP is associated with AE in both the pedigree individuals and the unrelated individuals (Note S4), and is also the SNP most strongly associated with usage of LD-based hotspots (P=1.8×10^{-52}) (Table S2). No locus outside PRDM9 is significant (P<0.01 after Bonferroni correction; Table S3). To better understand the association at rs6889665, we inferred the alleles in the PRDM9 ZF array carried by 139 individuals based on sequencing data from the 1000 Genomes Project^21, using the reads to infer each individual’s PRDM9 alleles among 29 alleles whose full sequences were previously determined^9 (Note S5). Grouping PRDM9 alleles based on how closely their binding target predictions match the 13-bp motif, following Berg et al.^9, we find that the ancestral “T” variant at rs6889665 is strongly correlated to “8/8 matches” to the 13-bp motif (including the “A” and “B” alleles), while the derived “C” variant is almost perfectly correlated to a group of “5/8 match” alleles, all predicted to bind a common, different, 17-bp motif “CCgCNgtNNNCgtNNCC”^9. This implies a common historical origin for alleles matching this 17-bp motif (Figure 2B; Figure S7; Note S5). We also experimentally measured the number of zinc fingers in PRDM9 in 354 individuals including 166 African Americans from the pedigree study (Methods). This showed, again, that rs6889665 differentiates PRDM9 alleles into two different classes, with 96% of haplotypes carrying the ancestral allele having ≤14 zinc fingers, and 93% of haplotypes carrying the derived allele having ≥14 zinc fingers (Figure S7). After conditioning on rs6889665, there is no evidence that ZF length is associated with the AE phenotype. Several SNPs near the PRDM9 ZF array show a conditional association signal that is much weaker than rs6889665, but still significant (Figure 2C; Figure S6; Note S4), with the strongest at rs10043097 (P=8.3×10^{-14}), upstream of the PRDM9 transcription start site. These SNPs may tag additional variation in PRDM9 ZF array, or potentially expression levels.

To directly identify candidate African-enriched hotspot motifs, we selected 2,454 loci with a high crossover rate in the AE Map and YRI Map (>2cM/Mb over 2kb), and no more than half this rate in the S Map and CEU Map (this set is more powerfully enriched for higher recombination in people of African ancestry than the 2,375 above, as it includes information from the contemporary maps). We compared these to a “control set” of 7,328 candidate hotspots more active in the European than the African derived maps (Methods; Note S6). To identify sequence motifs associated with the African-enriched hotspots^3,22, we identified short motifs that occurred at increased frequency in the African-enriched hotspot set (Note S6). Testing all motifs of length 5–9 bases revealed a 9-mer “CCCCAGTGA” (OR=1.79, P=2.24×10^{-8}, Bonferroni corrected P=0.004) which exhibited a kilobase-scale rate peak near occurrences of this motif in African derived maps, but in neither of the European derived maps (Figure S8). Further analysis revealed a strong influence of downstream flanking bases (Figure S9), and degeneracy, yielding a 17-bp consensus sequence “CCCCaGTGAGCGTtgCc” (Figure 3A; more strongly signaled bases are uppercase) with the same consensus obtained when we considered flanking sequence for only odd or even chromosomes, and whether we based the analysis on AE-S or YRI-CEU map comparisons (Note S6). The 500 best matches to this motif have a ~3-fold increase in average rate in the AA and YRI relative to the deCODE and CEU maps (Figure 3B, Figure S8). Hotspots...
associated with the motif occur in both unique and repetitive DNA (e.g. L1PA10/13 LINE elements; Figure S10) (Note S6). We also compared the 17-bp consensus to the binding motif predicted for “5/8 match” alleles, and found that they match almost precisely (Figure 3A; 10 of 11 bases, P=8.1×10⁻⁶).

How much of the African-enriched recombination pattern can be explained by PRDM9? We estimated the fraction of variation in the AE phenotype explained by rs6889665 in our pedigree data after accounting for noise in the phenotype estimation (Note S4). Over 82% of map usage variability is explained by rs6889665 genotype alone. Given there are further influential PRDM9 variants (Figure 3C), this gene may thus explain almost all differences in local rate between the West African and European populations. We next examined rates around 82 narrowly defined (<10kb) crossover sites in 7 individuals homozygous for the derived allele at rs6889665. There is no evidence of hotspots at these loci in either the deCODE or CEU Maps (Figure 3C), in contrast to events in individuals carrying the ancestral allele at rs6889665 (Figure S11). Thus, crossover positions in individuals who are homozygous for the derived allele at rs6889665 are consistent with an entirely different recombination hotspot landscape, which would imply PRDM9 control of all hotspots.

Despite the strong correlation between maps at megabase scales, there is mounting evidence that PRDM9’s influence on crossing-over may not be limited to fine scales: we observe a weakly significant association of rs6889665 with the total number of crossovers genome-wide in pedigrees (P=0.04), corresponding to an average 1.3 crossovers more per meiosis per derived allele, exceeding the strongest previously known association at RNF212.

We have shown that PRDM9 alleles that bind a novel 17-bp motif and occur at greatly increased frequency in people of West African ancestry have led to a shift in the recombination landscape compared with people of non-African ancestry. The larger number of hotspots available to West Africans implies that at the population level, crossovers are more evenly distributed than in Europeans, and thus the shorter extent of West African LD is not due to differences in demographic history alone (such as the lack of an out-of-Africa founder event). Our findings also have medical implications, as recombination errors leading to insertions or deletions are known to be associated with recombination hotspots. Our results predict that the congenital abnormalities that have been associated with the recombination hotspots bound by PRDM9 “A” and “B” alleles will occur at a decreased rate in people of West African ancestry, whereas new diseases will arise due to recombination errors near African-enriched hotspots.

**METHODS SUMMARY**

We assembled SNP array data from 29,589 unrelated people and 222 nuclear families genotyped at 490,000–910,000 SNPs from the Candidate Gene Association Resource (CARe), studies at the Children’s Hospital of Philadelphia (CHOP), the African American Breast Cancer Consortium, the African American Prostate Cancer Consortium and the African American Lung Cancer Consortium. To build a recombination map, we used HAPMIX to localize candidate crossover positions, and implemented a Markov Chain Monte Carlo (MCMC) that used the probability distributions for the positions of the filtered crossovers to infer recombination rates for each of 1.3 million inter-SNP intervals. We also
implemented a second MCMC that models each individual’s set of crossovers as a mixture of a Shared (S) Map similar to the European deCODE Map and an African-enriched (AE) Map, and then assigns each individual an “AE phenotype” corresponding to the proportion of their newly detected crossovers assigned to the AE Map. We imputed genotypes at up to three million HapMap2 SNPs using MaCH, and then tested each of these SNPs for association with the AE phenotype and other recombination-related phenotypes. We identified 2,454 candidate African-enriched hotspots with increased recombination rates in the YRI vs. CEU maps, and in the AE vs. S maps, and searched for motifs enriched at these loci, thus identifying a degenerate 17-bp motif. To study the structure of PRDM9, we measured the length of the PRDM9 zinc finger array and genotyped rs6889665 in YRI, CEU and the CARe nuclear families; we also carried out imputation based on 1000 Genomes Project short read data to infer the alleles individuals carry, among 29 previously characterized in a sequencing study of PRDM9.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

Authors

Anjali G. Hinch1, Arti Tandon2,3, Nick Patterson2, Yunli Song4, Nadin Rohland2,3, Cameron D. Palmer5,6, Gary K. Chen7, Kai Wang8,9, Sarah G. Buxbaum10, Meggie Akylbekova10,11, Melinda C. Aldrich12,13, Christine B. Ambrosone14, Christopher Amos15, Elisa V. Bandera16, Sonja I. Berndt17, Leslie Bernstein18, William J. Blot13,19, Cathryn H. Bock20, Eric Boerwinkle21, Quiyin Cai21, Neil Caporaso17, Graham Casey7, L. Adrienne Cupples22, Sandra L. Deming13, W. Ryan Diver23, Jasmin Divers24, Myriam Fornage25, Elizabeth M. Gillanders26, Joseph Glessner9, Curtis C. Harris27, Jennifer J. Hu28, Sue A. Ingles7, Williams Isaacs29, Esther M. John30, W. H. Linda Kao31, Brendan Keating9, Rick A. Kittles32, Laurence N. Kolonel33, Emma Larkin34, Loic Le Marchand33, Lorna H. McNeill35, Robert C. Millikan36, Adam Murphy37, Solomon Musani11, Christine Neslund-Dudas38, Sarah Nyante36, George J. Papanicolaou39, Michael F. Press37, Bruce M. Psaty40, Alex P. Reiner41, Stephen S. Rich42, Jorge L. Rodriguez-Gill28, Jerome I. Rotter43, Benjamin A. Rybicki38, Ann G. Schwartz20, Lisa B. Signorello13,19, Margaret Spitz15, Sara S. Strom44, Michael J. Thun23, Margaret A. Tucker17, Zhaoming Wang45, John K. Wiencke46, John S. Witte47, Margaret Wrensch46, Xifeng Wu15, Yuko Yamamura44, Krista A. Zanetti26,27, Wei Zheng13, Regina G. Ziegler17, Xiaofeng Zhu48, Susan Redline49, Joel N. Hirschhorn5,6,50, Brian E. Henderson7, Herman A. Taylor Jr.11,51,52, Alkes L. Price53, Hakon Hakonarson9,54, Stephen J. Chanock17, Christopher A. Haiman7, James G. Wilson55, David Reich2,3,* and Simon R. Myers1,4,*

Affiliations

1Wellcome Trust Centre for Human Genetics, Oxford University, Roosevelt Drive, Oxford OX3 7BN, UK
2Broad Institute of MIT and Harvard, 7 Cambridge Center, Cambridge, MA 02142, USA
3Dept. of Genetics, Harvard Medical School, New Research Bldg., 77 Ave. Louis Pasteur, Boston, MA 02115, USA
4Department of Statistics, Oxford University, 1 South Parks Road, Oxford OX1 3TG, UK
5Program in Medical and Population Genetics, Broad Institute, 7 Cambridge Center, Cambridge, MA 02142, USA
6Div. of Genetics & Endocrinology and Program in Genomics, Childrens Hospital Boston, MA 02115, USA
7Department of Preventive Medicine and Department of Pathology, Keck School of Medicine, University of Southern California/ Norris Comprehensive Cancer Center, Los Angeles, CA 90033, USA
8Zilkha Neurogenetic Institute, University of Southern California, Los Angeles, CA 90089
9Center for Applied Genomics, The Childrens Hospital of Philadelphia, Philadelphia, Pennsylvania 19104, USA.
10Jackson Heart Study Coordinating Center, Jackson State University, 350 W. Woodrow Wilson Ave., Suite 701, Jackson, MS 39213, USA
11Department of Medicine, University of Mississippi Medical Center, 2500 N. State St., Jackson, MS 39216, USA
12Department of Thoracic Surgery, Vanderbilt University School of Medicine, Nashville, TN 37203, USA
13Division of Epidemiology in the Department of Medicine, Vanderbilt Epidemiology Center; and the Vanderbilt-Ingram Cancer Center, Vanderbilt University School of Medicine, Nashville, TN 37203, USA
14Department of Cancer Prevention and Control, Roswell Park Cancer Institute, Buffalo, NY 14263, USA
15Department of Epidemiology, Division of Cancer Prevention and Population Sciences, The University of Texas MD Anderson Cancer Center, Houston, TX 7703
16The Cancer Institute of New Jersey, New Brunswick, NJ 08903, USA
17Division of Cancer Epidemiology and Genetics, National Cancer Institute, Bethesda, MD 20892, USA
18Division of Cancer Etiology, Dept. of Population Science, Beckman Research Inst., City of Hope, CA 91010, USA
19International Epidemiology Institute, Rockville, MD 20850, USA

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Karmanos Cancer Institute and Dept. of Oncology, Wayne State University of Medicine, Detroit, MI USA 48201

Human Genetics Center and Division of Epidemiology, University of Texas at Houston, 1200 Herman Pressler St., Houston, Texas 77030, USA

Department of Biostatistics, Boston University School of Public Health, 801 Massachusetts Avenue, Boston, MA 02118 and Framingham Heart Study, Framingham, MA 01702, USA

Epidemiology Research Program, American Cancer Society, Atlanta, GA 30303, USA

Department of Biostatistical Sciences, Wake Forest University School of Medicine WC-2326, Medical Center Blvd., Winston Salem, NC 27157, USA

Institute of Molecular Medicine and Division of Epidemiology, School of Public Health, University of Texas Health Sciences Center at Houston, 1825 Pressler Street, Houston, TX 77030, USA

Division of Cancer Control and Population Sciences, National Cancer Institute, Bethesda, MD 20892, USA

Laboratory of Human Carcinogenesis, Center for Cancer Research, National Cancer Institute, Bethesda, MD 20892, USA

Sylvester Comprehensive Cancer Center and Department of Epidemiology and Public Health, University of Miami Miller School of Medicine, Miami, FL 33136, USA

James Buchanan Brady Urological Institute, Johns Hopkins Hospital and Medical Institutions, Baltimore, MD 21287, USA

Cancer Prevention Institute of California, Fremont, CA 94538; and Stanford University School of Medicine and Stanford Cancer Center, Stanford, CA 94305, USA

Department of Epidemiology, Johns Hopkins Bloomberg School of Public Health, 615 N. Wolfe St., Baltimore, MD 21205, USA

Department of Medicine, University of Illinois at Chicago, Chicago, IL 60607, USA

Epidemiology Program, University of Hawaii Cancer Center, Honolulu, HI 96813, USA

Department of Medicine, Division of Allergy, Pulmonary and Critical Care, 6100 Medical Center East, Vanderbilt University Medical Center, Nashville, TN 37232-8300, USA

Department of Health Disparities Research, Division of OVP, Cancer Prevention and Population Sciences, and Center for Community Implementation and Dissemination Research, Duncan Family Institute, The University of Texas MD Anderson Cancer Center, Houston, TX 77030, USA
36Department of Epidemiology, Gillings School of Global Public Health, and Lineberger Comprehensive Cancer Center, University of North Carolina, Chapel Hill, NC 27599, USA

37Department of Urology, Northwestern University, Chicago, IL 60611, USA

38Department of Public Health Sciences, Henry Ford Hospital, Detroit, MI USA

39Division of Cardiovascular Sciences, National Heart, Lung and Blood Institute, 6701 Rockledge Drive, Bethesda, MD 20892, USA

40Cardiovascular Health Research Unit, Depts. of Medicine, Epidemiology & Health Services, Univ. of Washington; Group Health Research Institute; Group Health Cooperative; 1730 Minor Ave., Seattle, WA 98101, USA

41Department of Epidemiology, University of Washington, Box 357236 Seattle, WA 98195, USA

42Center for Public Health Genomics, University of Virginia, West Complex Room 6111, Charlottesville, VA 22908, USA

43Medical Genetics Institute, Cedars-Sinai Medical Center, 8700 Beverly Blvd, Los Angeles, CA 90048, USA

44Department of Epidemiology, The University of Texas M. D. Anderson Cancer Center, Houston, TX 77030, USA

45Core Genotype Facility, SAIC-Frederick, Inc., National Cancer Institute-Frederick, Frederick, Maryland, USA 20877, USA

46University of California San Francisco, San Francisco CA 94158, USA

47Institute for Human Genetics, Departments of Epidemiology and Biostatistics and Urology, University of California, San Francisco, San Francisco, CA 94158, USA

48Department of Epidemiology and Biostatistics, School of Medicine, Case Western Reserve University, Wolstein Research Building, Cleveland, Ohio 44106, USA

49Brigham and Women’s Hospital, Dept. of Medicine, Sleep Medicine, 75 Francis Street, Boston, MA 02115, USA

50Department of Medicine, Harvard Medical School, Boston, MA 02115, USA

51Jackson State University, 1400 Lynch Street, Jackson, MS 39217, USA

52Tougaloo College, 500 West County Line Road, Tougaloo, MS 39174, USA

53Departments of Epidemiology and Biostatistics, Harvard School of Public Health, Boston, MA 02115, USA

54Dept. of Pediatrics, University of Pennsylvania School of Medicine, Philadelphia, Pennsylvania 19104, USA.

55Department of Physiology and Biophysics, University of Mississippi Medical Center, 2500 N. State St., Jackson, MS 39216, USA
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Figure 1.
Building an African American genetic map. (A) HAPMIX detection of crossovers between segments of inferred ancestry is illustrated in a father-mother-child trio. Black segments show inferred crossovers; arrows show transmission of ancestral crossovers from parent to child. Purple/green segments show de novo events (paternal/maternal origin respectively), corresponding to events identified directly using two additional children (bottom, "Pedigree inferred"). (B) The AA Map localizes five hotspots in a region of the MHC whose positions (blue) were previously mapped by sperm typing\(^1\). (C) Comparison of maps shows a hotspot at 33.1Mb in the African-derived AA and YRI maps, but not the deCODE and CEU maps (all maps smoothed to 10kb).
Figure 2.
Association of \textit{PRDM9} genetic variation with hotspot activity. (A) A GWAS measuring association of the “African-enrichment” (AE) phenotype shows a single genome-wide significant peak at \textit{PRDM9}, with rs6889665 the best associated SNP. (B) Relationship between alleles at the rs6889665 and predicted binding target of the \textit{PRDM9} zinc finger array\textsuperscript{9} for West African and European samples. The alleles are grouped into 8 clusters according to their best-matching region to the 13-bp motif, and annotated by the number of bases matching the motif. The African-enriched rs6889665 “C” allele always co-occurs with
motifs with a poor (5/8) match to the 13-mer. (C) Gene tree of the LD block containing the PRDM9 ZF array (Methods); numbered circles show SNPs and significant P-values for association, after conditioning on rs6889665.
Figure 3.
A sequence motif specifying the positions of African-enriched hotspots. (A) Logo plot showing a degenerate 17-bp hotspot motif, with stack height proportional to -log P-value, and relative letter height proportional to the mean crossover rate increase given each base. Below is the bioinformatic PRDM9 binding prediction for the rs6889665 AE associated alleles (from Figure 2B), matching the motif at 10/11 bases (lines). (B) Average crossover rate (in 2 kb sliding windows) in the AA (red line) and deCODE (black line) maps surrounding the 500 strongest motif matches. (C) In seven rs6889665 “CC” individuals from the pedigree study, we localized 82 crossovers to within 10 kb, and plot average AA, YRI, deCODE and CEU map rates. There is no strong peak above local background in the deCODE or CEU maps.
Table 1

Genetic map assessments at different size scales

| Scale (interval size) | \( \rho \) - Pearson correlation of the AA map (deCODE Map) to the specified LD map | Est. correlation of AA Map to the true map (inferred by MCMC)* | Est. coefficient of variation of AA Map (std. err. divided by crossover rate expected for interval size)** |
|----------------------|---------------------------------------------------------------------|---------------------------------------------------------------|--------------------------------------------------------------------------------------------------|
|                      | Combined LD§ | CEU (0.58) | YRI (0.74) |                                                                  |                                                                 |
| 3 kb                 | 0.75 (0.63) | 0.66 (0.58) | 0.71 (0.53) | 0.93                                                                 | 1.41 |
| 10 kb                | 0.82 (0.74) | 0.73 (0.70) | 0.78 (0.65) | 0.96                                                                 | 0.73 |
| 30 kb                | 0.86 (0.83) | 0.78 (0.78) | 0.83 (0.74) | 0.98                                                                 | 0.36 |
| 100 kb               | 0.91 (0.89) | 0.84 (0.85) | 0.87 (0.81) | 0.99                                                                 | 0.17 |
| 300 kb               | 0.94 (0.93) | 0.89 (0.90) | 0.92 (0.88) | 1.00                                                                 | 0.08 |
| 1 Mb                 | 0.97 (0.96) | 0.94 (0.94) | 0.95 (0.95) | 1.00                                                                 | 0.04 |
| 3 Mb                 | 0.98 (0.98) | 0.97 (0.97) | 0.98 (0.97) | 1.00                                                                 | 0.02 |

Note: The numbers in this table are restricted to the autosomes and genomic segments more than 5 Mb from the telomere.

§ The Combined map is the HapMap2 population-averaged LD-based map18.

* The standard error of the map at each size scale is determined by the posterior probability distribution of the MCMC.

** The Combined LD map is the HapMap2 population-averaged LD-based map18.