KinSNP software for homozygosity mapping of disease genes using SNP microarrays

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Abstract

Consanguineous families affected with a recessive genetic disease caused by homozygotisation of a mutation offer a unique advantage for positional cloning of rare diseases. Homozygosity mapping of patient genotypes is a powerful technique for the identification of the genomic locus harbouring the causing mutation. This strategy relies on the observation that in these patients a large region spanning the disease locus is also homozygous with high probability. The high marker density in single nucleotide polymorphism (SNP) arrays is extremely advantageous for homozygosity mapping.

We present KinSNP, a user-friendly software tool for homozygosity mapping using SNP arrays. The software searches for stretches of SNPs which are homozygous to the same allele in all ascertained sick individuals. User-specified parameters control the number of allowed genotyping 'errors' within homozygous blocks. Candidate disease regions are then reported in a detailed, coloured Excel file, along with genotypes of family members and healthy controls. An interactive genome browser has been included which shows homozygous blocks, individual genotypes, genes and further annotations along the chromosomes, with zooming and scrolling capabilities. The software has been used to identify the location of a mutated gene causing insensitivity to pain in a large Bedouin family. KinSNP is freely available from http://bioinfo.bgu.ac.il/bsu/software/kinSNP.

Keywords: autozygosity, consanguinity, genetic mapping, homozygosity, SNP microarrays, software

Introduction

The availability of high-density mapping microarrays, bearing sufficient probes to analyse 10,000–1,000,000 single nucleotide polymorphisms (SNPs) in one assay, offers an efficient alternative to traditional, microsatellite-based, genome-wide linkage scans. Accordingly, several public linkage software programs have been developed, such as Allegro, EasyLinkage and dChip, which can handle the high number of SNPs in these arrays. All of these have

been successfully used for mapping Mendelian disorders.^{2–4}

Genotyping individuals from large, multiply-consanguineous families of isolated populations offers a unique advantage for positional cloning of rare diseases. ^{5,6} A mutation occurring in a founder may be rapidly inherited by numerous individuals in the population, and the offspring of consanguineous parents will have a high probability of inheriting two copies of the mutated chromosomal segment and thus expressing the disease. Classical

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linkage analysis of SNP arrays in these studies is highly problematic, however, due to the computational load needed to deal with both the large number of SNPs and the family structure. The presence of complex inbreeding loops makes it extremely difficult to reconstruct genetic data from untyped individuals. Current linkage programs designed to deal with this challenge (eg SuperLink⁷ and SimWalk2⁸) are not suited for the large number of markers present in SNP arrays. The latter programs may be used if adjacent SNPs are grouped and regarded as one marker (eg Naveed et al. ⁹); however, this causes great loss of the information available in the high-density array.

A powerful alternative to linkage analysis, which has long been used in studies of autosomal recessive diseases in consanguineous pedigrees, is homozygosity (or autozygosity) mapping. 10,11 This strategy relies on the fact that in a child of a consanguineous marriage affected with a homozygous recessive genetic disease mutation, a large region spanning the disease locus is probably also homozygous. Thus, searching for regions that are consistently homozygous by descent in all affected individuals would be expected to provide a suitable strategy for mapping the recessive gene mutation. Moreover, the use of SNP arrays is extremely advantageous for homozygosity mapping, compared with microsatellites, because of the much higher density of markers they comprise. For economic reasons, a common strategy is to perform homozygosity mapping using SNP arrays for a small number of sick individuals, with or without their healthy parents or siblings. Once candidate regions are found, they are then further analysed using microsatellite markers in a large number of family members, where all of the affected and none of the unaffected members should be homozygous to the same allele. 12-19 In the future, SNP typing or targeted sequencing could be used for fine-mapping. Subsequently, linkage logarithm of odds (LOD) scores are calculated using genetic programs such as SuperLink or SimWalk2.

The ever-increasing number of markers in the SNP arrays makes it unfeasible to locate homozygous regions by manual inspection of the genotype

file, as used to be done with microsatellite data. At the time of writing, only two public software programs, ExcludeAR²⁰ and AutoSNPa,²¹ were available for automated homozygosity analysis of SNP arrays.

We present KinSNP, a new software tool for homozygosity mapping in SNP arrays from consanguineous families, which provides additional capabilities compared with existing software while simplifying program usage. KinSNP can be run on any computer through intuitive menus and dialogues. It directly accepts Affymetrix files as input, whereas files from Illumina or other platforms may be easily converted to the KinSNP format. The user has considerable control over the parameters that define the reported homozygous regions, allowing, for example, various degrees of stringency for tolerating genotyping errors. KinSNP results are provided both as coloured, linked Excel files and as an interactive genome browser, on which results from multiple analyses may be visualised simultaneously.

In the present study, KinSNP software was used to identify the location of the mutated gene causing insensitivity to pain (IP) in a large Bedouin family.

Results

Outline of KinSNP

KinSNP was developed to identify runs of shared homozygosity (ROSHs) in SNP microarrays. The software was initially designed to mimic the dataprocessing strategy traditionally used by geneticists manually locate homozygous regions microsatellite-based data. One of our first microarraybased studies, aimed at mapping IP in an inbred Bedouin family using four Affymetrix 250K Nsp mapping arrays (Figure 1), revealed a 16-megabase (Mb)-long ROSH. The ROSH was further confirmed by typing a larger set of family members using microsatellites, and the causative mutation was identified through sequencing in an already known IP gene, SCN9A.²² We use that analysis here to demonstrate KinSNP operation and output. Since only one healthy relative was genotyped in the study, PRIMARY RESEARCH Amir et al.

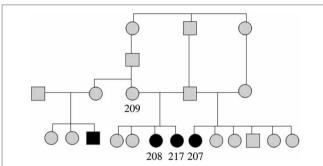


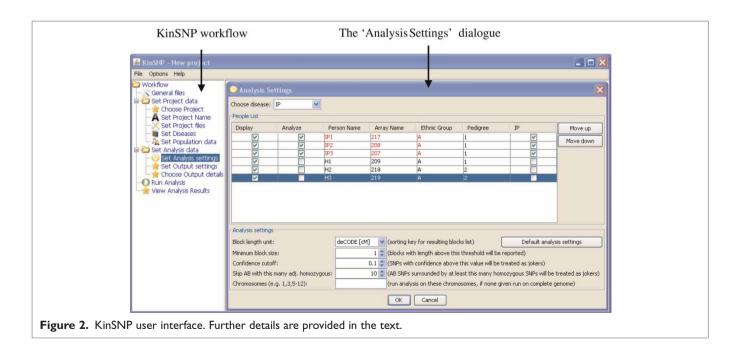
Figure 1. Insensitivity to pain (IP) family tree. The numbered individuals were genotyped using Affymetrix SNP microarrays.

individuals from the same population studied for other diseases in our group were added as healthy controls for IP. This inspired us to enable KinSNP to receive all of a laboratory's or a core facility's microarray genotypes at once, and be subsequently used iteratively to look for homozygous regions in any selected set of sick individuals with any combination of analysis parameters. Program execution is carried out through an intuitive guide, which leads the user through the analysis workflow (Figure 2, left-hand panel). Eventually, the candidate disease regions are provided both as a textual, sorted list (Figure 3), as well as a visual representation in the context of the entire genome (Figures 4, 5).

KinSNP analysis workflow

The analysis of microarray data often requires stepwise formatting and combining of multiple files in various formats. To simplify this step, KinSNP requires only two input files: (i) an array-specific annotation file, containing SNP identifications accompanied by their genetic and physical positions and (ii) a genotype file, containing SNP genotype and confidence calls from all typed individuals. Instructions on the exact format of these simple tab-delimited text files are available on the KinSNP website. KinSNP is also able to directly accept files produced by Affymetrix genotype calling programs (GTYPE, Genotyping Console), as well as Affymetrix annotation files (downloadable from http://www.affymetrix.com/support/technical/ byproduct.affx?cat=software). Illumina data may be converted for use by KinSNP through a dedicated script available at the KinSNP website.

Once the input files are loaded into KinSNP, the user is allowed to associate a name, a disease state, a family and an ethnic group to each array included in the genotype file. This step of defining the 'population' accomplishes the one-time 'project' description, after which the user may perform any number of specific homozygosity analyses on any of the diseases and subsets of individuals.



The parameters for the homozygosity analysis are specified in the 'Analysis Settings' window shown in detail in Figure 2. The user should specify the studied disease, choose the specific chips (individuals) in which to look for shared homozygous regions and denote which other individuals to show in the output as controls for manual inspection. The reported ROSHs can then be sorted according to one of several possible length units, such as number of SNPs, physical length and genetic recombination length according to deCODE, Marshfield or SNP Linkage Map (SLM1) genetic maps. (Information about these maps and their extrapolation to obtain SNP positions is available in the README file that accompanies Affymetrix annotation files.) addition, the user can specify a length threshold above which ROSHs will be reported because this may differ between homozygosity projects. The more related the individuals and the fewer there are, the longer the ROSHs that can be expected to be found. For example, brothers are expected to share longer ROSHs than cousins; a longer ROSH is expected to be shared by two patients than by three patients etc. Additionally, the more recent the mutation and the higher the consanguinity, the longer the ROSHs expected. The parameters 'Confidence cutoff' and 'Skip AB with this many adj. homozygous' control the number of allowed errors within ROSHs and are explained in the next section. For faster computation, the user may limit the analysis to specific chromosomes only.

After pressing 'OK', the user is asked to specify output file names and locations, and analysis may be started.

Homozygosity analysis algorithm

First, KinSNP merges positional information from the annotation file with the genotype data and sorts the genotypes by SNP physical locations. Starting with the first SNP, the algorithm checks whether all sick individuals are homozygous to the same allele; if not, it proceeds to the next SNP. If the individuals share homozygosity, a new ROSH is started. The ROSH is extended as long as homozygosity is shared. Under the most stringent

parameters, once a heterozygous SNP or a non-shared homozygous SNP for the other allele are encountered, the ROSH is terminated and stored. The algorithm then continues to proceed along the chromosome until it encounters another SNP with shared homozygosity. It then starts a new ROSH and so forth, until the end of the chromosome is reached.

A more permissive operation of KinSNP allows for occasional non-shared alleles within ROSHs. The degree of permissiveness may be heuristically adjusted via two parameters in the KinSNP Analysis Settings window, which take into account the call confidence and the scarcity of the non-shared allele.

The first parameter, 'Confidence cutoff', makes use of the confidence values attributed to each of the genotype calls. To understand this parameter, let us first explain the 'NoCall' genotypes. By default, SNP calls designated as 'NoCall' in the KinSNP input file are ignored during the process of ROSH identification. Thus, a 'NoCall' at a certain SNP will be treated as AA or as BB if this will cause the SNP to be considered homozygous among all individuals selected for analysis. The designation of SNPs as 'NoCalls' is usually applied by the genotype calling program (eg Affymetrix's Genotyping Console) if the SNP confidence value did not pass a certain cut-off. KinSNP enables the user to change that cut-off by specifying another confidence threshold above which SNP calls will be effectively regarded as 'NoCalls' in ROSH definition. Unlike 'NoCalls', however, these SNPs will be shown in KinSNP Excel output files with their original genotype call, though coloured with a lighter colour than the others (see for example Figure 4).

The second parameter, 'Skip AB with this many adjacent homozygous', was introduced as we found that heterozygous SNPs — even with high confidence calls — sometimes 'break' genuine homozygous blocks, probably due to erroneous reading. This parameter enables one to ignore heterozygous SNP calls which are surrounded by a 'window' of a user-specified number of SNPs which all apply to the criteria of the ROSH.

From our experience, the two parameters, as well as the different colouring of low-confidence

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Chr	Block name	StartPos	EndPos	No. of SNPs	Phys Length	deCODE	Marshf	SLM1
2	chr2.6	153,087,633	169,273,511	1455	16,185,878	11.5793	17.5383	13.8066
9	chr9.2	38,736,473	70,359,435	28	31,622,962	4.5823	3.3384	0.1486
15	chr15.2	26,187,876	27,528,814	51	1,340,938	3.8995	2.8899	2.3547
3	chr3.1	7,960,207	9,069,531	154	1,109,324	3.4535	4.8174	3.9069
22	chr22.4	23,535,719	24,477,557	68	941,838	3.3888	2.6540	3.0760
9	chr9.7	136,577,004	137,166,820	33	589,816	3.2839	3.9039	2.6080
2	chr2.5	134,260,660	138,342,588	430	4,081,928	3.2709	3.5438	6.2698
19	chr19.13	60,374,789	61,011,337	21	636,548	2.3122	1.5743	3.6913
17	chr17.8	74,995,271	75,313,177	2	317,906	2.2862	3.4761	0.8213

Figure 3. Excel output: Summary of the ROSHs fitting the user definition. ROSH ('block') positions are indicated in number of bases from pter. ROSH ('block') names are hyperlinks to block start positions in the chromosome files (Figure 4). Block lengths are indicated as number of SNPs, physical length (number of bases), and deCODE, Marshield and SLMI genetic distances (cM).

genotypes, not only improve the homozygosity analysis, but also enable inclusion in the analysis of low-quality arrays (eg <90 per cent call rate) which otherwise could not be analysed.

Viewing KinSNP results

KinSNP output may be viewed either as a set of stylised Excel files or in an interactive genome browser. The Excel files enable the user to further sort, mark and annotate the results using the standard Excel interface. The browser enables scrolling and zooming along the chromosomes, thus visually identifying recombination patterns flanking the reported ROSHs and estimating their likelihood as disease-related regions. We will demonstrate KinSNP output by describing the IP disease gene mapping.

Excel output

KinSNP creates a summary file which outlines all the found ROSHs longer than the user-specified threshold, along with their chromosomal position, number of consecutive SNPs and physical and genetic lengths (Figure 3). After sorting the ROSHs according to the desirable length unit, the IP analysis revealed a ~11–17 centimorgan (cM) (~16 Mb)-long block comprising 1,455 consecutive SNPs in chromosome 2. Clicking the active link in the block name opens the chromosomespecific Excel file (Figure 4), right at the top of the relevant block. In this file, genotypes of all SNPs of the chromosome are shown in the order of the SNPs' physical positions, accompanied by the SNP annotation retrieved from the annotation input file.

Genotype calls from the selected sick individuals appear first, with reported ROSHs delineated by black vertical bars. Genotypes of control individuals are shown further to the right. Heterozygous SNPs are coloured in green and homozygous SNPs are coloured in red (AA) or blue (BB), respectively, such that the allele in lower alphabetical order is the A allele and the other allele is the B allele. Low-confidence calls, based on user-specified confidence cut-off, are coloured in light green, orange or light blue, for AB, AA or BB, respectively.

The Excel display allows the user to judge how informative the ROSHs are by comparing the degree of homozygosity of the ROSHs in the sick individuals with the degree of heterozygosity in the control individuals. SNPs or blocks which are homozygous across both sick and healthy individuals should be considered non-informative. Note bordering non-ROSH SNP_A-1821170 in Figure 3) is included at each end of the reported ROSH, since the disease-causing mutation may in principle be located between that SNP and the extreme ROSH SNP. Users may also scroll above and below the reported ROSHs to see whether several adjacent ROSHs can actually be merged, and to get an impression of the recombination patterns flanking the ROSHs. These tasks are more easily performed using the included Integrated Genome Browser (IGB), however.

The Integrated Genome Browser

In the IP analysis shown here, one ROSH was found to stand out from all the others, being much

SNP ID	rs ID	Chr	Physical pos	deCODE	marshfield	SLM1	IP1	IP2	IP3	H1	H2	H3
SNP_A-2021017	rs13427485	2	153,086,232	162.2921	155.2904	149.7633	88	88	AB	NO_CAL	AB	AB
SNP A-4207737	rs4664114	2	153,086,705	162.2924	155.2911	149.7635	AA	AA	AB	NO_CAL	AB	AB
SNP A-4223103	rs4664593	2	153,086,898	162.2925	155.2913	149.7636	AA	AA	AB	AA	AB	AB
SNP A-1821170	rs16831279	2	153,087,633	162.2930	155.2924	149.7639	BB		AB	* 88	AB	
SNP A-4226014	rs16831280	2	153,087,773	162.2931	155.2926	149.7640				BB	NO_C	ILL BB
SNP A-2061026	rs2678298	2	153,100,954	162.3010	155.3115	149.7696	AA	AA	AA	* AA	AA	AA
SNP A-2179521	rs2577181	2	153,102,012	162.3016	155.3131	149.7700	BB	88		- AB	AB	25
SNP A-1905185	rs2577180	2	153,102,193	162.3017	155.3133	149.7701			AB	BB	AB	55
SNP A-1965413	rs10497108	2	153,146,684	162.3283	155.3772	149.7889				BB	AB	AB
SNP A-1965414	rs16823807	2	153,146,906	162.3285	155.3775	149.7890	AA	AA	AA	* AA	AB	AB
SNP A-1965415	rs10497110	2	153,146,937	162.3285	155.3776	149.7890	AA	AA	AA	• AA	AA	AA
SNP A-4208425	rs16823809	2	153,147,303	162.3287	155.3781	149.7892				* 88	AB	AB
SNP_A-4208426	rs10497111	2	153,147,664	162.3289	155.3786	149.7893				BB	AB	AB
SNP_A-1942192	rs7581918	2	153,154,459	162.3330	155.3884	149.7922				BB		AB
SNP_A-1781652	rs6710689	2	153,156,492	162.3342	155.3913	149.7931						
SNP_A-1965419	rs12693415	2	153,181,767	162.3493	155.4276	149.8062				* AB	AB	AA
SNP_A-1965420	rs2304556	2	153,182,040	162.3495	155.4280	149.8064				AB	AB	AA
SNP_A-1851689	rs10178618	2	153,195,011	162.3573	155.4466	149.8153	AA.	AA	AA	* AB	AA	AA
SNP_A-2194475	rs4664599	2	153,201,011	162.3609	155.4552	149.8195				N 88	88	
SNP_A-2263370	rs1561267	2	153,201,156	162.3609	155.4554	149.8196	AA	AA	AA	* AB	AA	AA
SNP_A-2024466	rs2272535	2	153,201,532	162.3612	155.4560	149.8198				* AB		
SNP_A-1965422	rs6738344	2	153,206,156	162.3639	155.4626	149.8230					AB	
SNP_A-2281626	rs3171980	2	153,231,102	162.3789	155.4985	149.8402	AA	AA	AA	* AB	AB	AA
SNP_A-2055374	rs3762503	2	153,231,209	162.3789	155.4986	149.8403	AA	AA	AA	* AA	AB	
SNP_A-2054640	rs2083097	2	153,233,532	162.3803	155.5019	149.8419	AA	AA	AA	* AB	BB	
SNP_A-2028616	rs10181036	2	153,235,937	162.3818	155.5054	149.8436				* BB		
SNP_A-2095815	rs16831742	2	153,322,377	162.4335	155.6296	149.9032	AA	AA	ДД	* AA	AA	AA
SNP_A-2157738	rs13036075	2	153,339,251	162.4436	155.6538	149.9149	AA	AA	AA	NO_CAL	AB	AB
SNP_A-1781390	rs11883698	2	153,340,083	162.4441	155.6550	149.9154				1 33	BB	
SNP_A-2127334	rs2114653	2	153,356,833	162.4542	155.6790	149.9270	AA	AA	AA	* AA	AA	AA
SNP_A-2163059	rs10432430	2	153,356,905	162.4542	155.6791	149.9271						
SNP_A-1920668	rs17401154	2	153,391,842	162.4751	155.7293	149.9512	AA	AA	AA	* AB	AA	AA
SNP_A-2154496	rs13025951	2	153,421,299	162.4928	155.7716	149.9715					AB	AB

Figure 4. Excel output: Chromosome view of genotypes and ROSHs. Shown is the start of the 16-Mb-long ROSH found in chromosome 2. Genotype calls having lower confidence (higher values) than the user-specified cut-off are displayed with a lighter colour.

longer in both physical and genetic length, as well as having a greater number of SNPs. When there are several possible ROSHs, however, additional means are required to prioritise the ROSHs as candidate regions harbouring the disease-causing mutation. One way is to consider the degree of kinship among the ascertained individuals, and to check whether closer relatives indeed share longer ROSHs than those shared by all patients together. This is achieved by providing an informative visualisation of the ROSHs, along with genotypes of sick and control individuals, in a more compact and yet resizable view of the chromosomes. To this end, we have integrated with KinSNP a slightly simplified version of Affymetrix's IGB (http://www.affymetrix. com/support/developer/tools/download_igb.affx), whose source code is covered by the common public licence. By choosing KinSNP's 'View Analysis Results' option (Figure 2, left-hand panel), the IGB opens right away, with an option to add

tracks for the reported blocks and for coloured genotypes of the analysed individuals (Figure 5). University of California Santa Cruz (UCSC) gene predictions based on data from RefSeq, Genbank and UniProt are displayed by default, but additional genome annotations may be easily loaded to IGB from a variety of public and private databases, including, for example, RefSeq and Genbank mRNAs, GenScan gene predictions, microRNAs, Affymetrix probes from various array types, etc. The major advantage of the IGB for viewing KinSNP results is the possibility of navigating easily along the chromosomes and of zooming in and out of areas of interest. For example, in the IP ROSH hosting the SCN9A gene (Figure 5), it is possible to see the longer ROSH shared by the sisters IP1 (217) and IP2 (208) (on the left-hand side), overlapping with the shorter homozygous stretch shared with their half-sister IP3 (207). Once again, although the present KinSNP computed analysis does not

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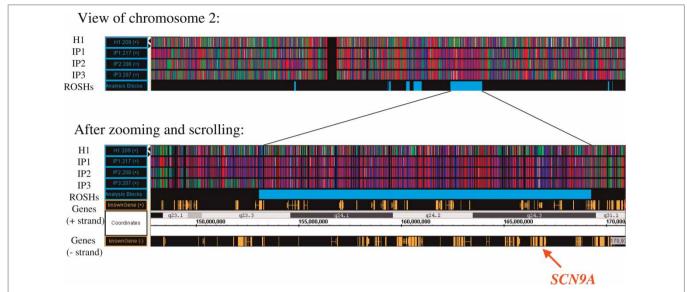


Figure 5. IGB view of KinSNP results. The IP causative gene, SCN9A, is marked in red. Note that the sisters indicated as IP1 and IP2 share a longer ROSH than their half-sister IP3, whereas their mother, H1, is heterozygous along the region.

consider the family tree in prioritising the reported blocks, the user may use knowledge of kinship between the ascertained individuals in combination with the IGB visualisation to estimate the reliability of each of the reported ROSHs.

The information which IGB uses to draw the block positions (Figure 5, 'ROSHs' track, shown as light blue rectangles) is read from a special text file created by KinSNP and stored in the same local directory as the Excel files. This file has a standard format, called a general feature format (GFF), and is recognised by its '.gff' extension. Once the IGB is open (often with the results of the most recent analysis), it is possible to load on top of it additional .gff files created in previous analyses. For example, if patients from two families were initially analysed separately and then all together, or if closely related patients were analysed and then more distant ones were added, computed blocks from all analyses may be displayed together in the same IGB view and compared visually. This option is also relevant when KinSNP is executed iteratively with different analysis parameters, such as minimal block lengths or confidence cut-off.

Finally, it is possible to load computational results from other genome-related experiments, expressed as binary or numerical values attributed to individual genomic positions or intervals, into the IGB on top of any existing display. These may include, for example, gene expression data, copy number variation data, measures of conservation between different genomes, etc. This allows straightforward integration of homozygosity analysis results with related custom or public experimental data.

Discussion

KinSNP software greatly facilitates homozygosity analysis of high-throughput SNP genotypes from consanguineous families, and may serve as an efficient starting point for further microsatellite-based fine mapping of candidate regions using classical linkage analysis. Moreover, the possibility of identifying homozygous blocks could be useful for the identification of a *de novo* mutation occurring in a recent generation in the general population. In such a case, homozygotisation of the mutation may occur in a patient owing to a founder effect, even if kinship is unknown to the family.

Unlike the two existing equivalent software tools, ExcludeAR and AutoSNPa, reported ROSHs are defined by customised criteria, which consider genotype call confidence and allow for

sporadic heterozygotes within specified-length windows of shared homozygosity. Additionally, block sizes are specified by physical length, distances according to several genetic maps and number of SNPs. Data import is easy and, for Affymetrix arrays, does not require any prior manipulation of the genotype or annotation files. KinSNP output presents the ROSHs within the context of entire chromosomes, using Excel files as well as an interactive genome browser. In future versions of KinSNP, we plan to provide additional visualisations — such as continuous plots along the chromosomes designating the degree of shared homozygosity in each position, and its comparison with that of the healthy controls.

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