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Structural and Genomic Variation in Preterm Birth

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Abstract

Background—Runs of homozygosity (ROH) are consecutive homozygous genotypes, which may result from population inbreeding or consanguineous marriages. ROH enhance the expression of recessive traits.

Methods—We mapped ROH in a case control study of women delivering at term compared to women delivering at or before 34 weeks gestation. Gene sets known to be important in risk of preterm birth were examined for their overlap with identified ROH segments.

Results—While we found no evidence of increased burden of ROH or copy number variations in mothers delivering at or before 34 weeks compared to term, we identified 424 genome-wide 50 kb segments with significant difference in abundance of overlapping ROH segments in cases versus controls, p<0.05. These regions overlap 199 known genes. We found preterm birth associated genes (CXCR4, MYLK, PAK1) and genes shown to have an evolutionary link to preterm (CXCR4, PPP3CB, C6orf57, DUSP13, and SLC25A45) with significant differences in abundance of overlapping ROH blocks in cases versus controls, p<0.001.

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Data Access

The results are based on SNP genotype data from the GENEVA study of the Danish National Birth Cohort, available in dbGaP PHS 000103.v1.p1. The genes used for the analysis are based on the Database for Preterm Birth(41).

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Conclusion—We conclude, while we found no significant burden of runs of homozygosity, we did identify genomic regions with significantly greater abundance of ROH blocks in women delivering preterm that overlapped genes known to be involved in preterm birth.

INTRODUCTION

Human genetic variation contributes to our individual characteristics, our susceptibility to disease and our individual responses to medications. This genetic variation can take the form of single nucleotide polymorphisms, copy number variations, chromosomal duplications and chromosomal loss. Runs of homozygosity (ROH) are an important part of genetic variation(1). ROH are stretches of consecutive homozygous genotypes, which may result from identity by descent, population inbreeding and/or cultural support for consanguineous marriages. Abundant ROH enhance the expression of recessive traits. Evolutionary pressures and purifying selection purge deleterious variants, but elevate the frequency of haplo-types surrounding a favored allele. Recent reports have shown that long runs of homozygosity are enriched for deleterious variants(2). This is especially true for ROH that span several megabases. The ability to do homozygosity mapping through examination of ROH has been enhanced by the availability of high-density genotyping arrays and has contributed to our understanding of homozygosity in both inbred and outbred populations. Reports using this approach in large numbers of individuals with very low evidence for inbreeding (up to 10 generations) have shown that ROH up to 4 Mb is common(1, 2).

Despite significant advances in the care of low birth weight infants, preterm birth remains a leading cause of newborn morbidity, mortality and hospitalization in the first year of life(3). Moreover, despite numerous attempts at intervention, the incidence of prematurity has shown minimal improvement over the last two decades. The risk factors associated with prematurity are many, including: adverse socio-demographic factors, race/ethnicity, infection, stress, trauma and prior history of a premature birth. The leading etiology is idiopathic. A large number of clinical/epidemiologic studies have examined the individual and collective contribution of each of these factors(4–6). There is substantial evidence for a genetic contribution to the risk of preterm birth(4–11). Twin studies suggest heritability is 36–40%, however differences in gestational age used and other details cloud the precision of those estimates (9, 10). Large epidemiological analyses drawn from population based studies support a maternal origin for the genetic contribution(s) to risk of preterm birth, with little contribution by paternal or fetal genetic factors(7, 12–14). A family history of preterm birth and inter-pregnancy interval of <18 months also increase the risk of prematurity (7, 10, 15). Population-based studies of consanguinity have suggested that there may be an increased risk of autosomal recessive inheritance associated with preterm birth (16, 17). One study found cousin marriage to result in a 1.6-fold increase risk of spontaneous preterm birth before 33 weeks gestation(16). A similar study found an odds ratio of 1.5 (CI, 1.2–1.9) of increased risk of preterm birth when mother and father are related (17). Whether one or several of these genetic etiologies converge via a final common pathway is unclear.

In an attempt to better understand the contribution of structural and genomic variation to the genetic etiology of preterm birth, we analyzed a large genome wide association study of preterm birth for structural variations (copy number variations, insertions and deletions) and

for large genomic stretches of runs of homozygosity. We analyzed high-density genotyping data from the gene environment association studies initiative (GENEVA) funded by the trans-NIH Genes Environment and Health Initiative. The GENEVA study included 4,000 Danish women and children from the Danish National Birth Cohort, for which phenotype and genotype information from a genome-wide association study are available. In order to study a more parsimonious group of preterm births which were more likely to be due to genetic disorders than environmental causes, we restricted our analysis to mothers that delivered at less than 34 weeks gestation. Thus, we focused on the genotype data from mothers who delivered at term (n=1018) and those who delivered prematurely (<34 weeks gestation n=454). We found no association with copy number variations and preterm birth. While we saw no evidence of significant burden of runs of homozygosity in preterm birth, there are significant genomic regions with ROH associated with genes known to be in involved in preterm birth.

RESULTS

Geneva Dataset

This study is a secondary analysis of genotype data from a genome-wide study of preterm birth known as Genome-Wide Association Studies of Prematurity and Its Complications, dbGaP Study Accession: phs000103.v1.p1. All phenotypic and genotype information deposited into dbGAP and were made available to us. The data represent a genome-wide case/control study using approximately 1,000 preterm mother-child pairs from the Danish National Birth Cohort (DNBC) along with 1,000 control pairs where the child was born at ~40 weeks' gestation. This study employed whole genome genotyping, on the ILLUMINA Human 660W-Quad_v1_A array. No significant univariate genome-wide associations were found related to preterm birth and this data is available as supplementary material (see Supplementary File S1 online). We used the genotype data from the mothers who delivered at term (n=1018) and those who delivered prematurely (<34 weeks gestation n=454) to carry out homozygosity mapping.

Population Structure and Relatedness

Principal component and relatedness analysis of the original Geneva dataset were employed to confirm that any identified associations were not due to unrecognized family structure with the dataset and this data is available as supplementary material (see Supplementary File S1 online). The relatedness estimates identified one unexpected duplicate that involved a misidentified gender subject, which was been removed from the data set. Also unexpected were two half-sib-like pairs. Otherwise 3 full siblings, 2 half siblings, and 10 first cousins were detected. For each pair of the families associated with one of the two unexpected half-sib-like pairs, only one mother was picked to be included in further analyses. The population structure using principal components analysis (PCA) was conducted on the original 3,867 mother-child subjects from the GENEVA project, along with HapMap anchor controls (CEU, YRI, CHB and JPT). All of the study samples cluster near the CEU HapMap anchor point as expected and this data is available as supplementary material (see Supplementary Figure S2 online). Because of the inclusion criteria requirement that the infant's parents and

all four grandparents must be of Danish origin, no additional analysis of within population stratification was conducted.

Structural Variation and CNV Analysis

We examined the data for univariate association of CNVs with preterm birth and to adjust for the spurious appearance of ROH due to regions of hemizygosity. CNV analysis was performed with the Illumina Bead Studio Program and PennCNV(18). Using PLINK to perform association testing on segmental CNV data from the PennCNV variants, we did not identify any significant segregation of CNVs (indels, duplications) between the preterm birth cases and the term controls. We did identify a total of 90 hemizygous CNVs from 78 mothers that were used to correct for spurious ROH. We also found that 16 of the ROH blocks (minimum segment of 1000kb or larger) overlapped with CNV blocks.

Homozygosity Burden

The burden of homozygosity was analyzed for three categories of minimum final segment length (1000 kb, 2000 kb, 3000 kb). For each minimum segment length, the proportion of individuals with at least one ROH, the number of ROHs per individual and the total length ROHs per individual were compared between the case and control cohorts, with case-control status determined by gestational age (GA). Control (individuals with GA greater or equal to 38 weeks) vs case (individuals with GA less than or equal to 34 weeks) status was analyzed with logistic regression using differences in the proportion, total number and total length of ROH as predictors and maternal age as a covariate. The total lengths of ROHs per individual were also calculated and the means are reported for four mutually exclusive gestational age categories.

We did not find a statistically significant association between the proportion of individuals with at least one ROH and case/control status for any of the ROH minimum length categories (Table 1). Additionally, when considering only individuals with at least one minimum ROH length segment, while the average number of ROH per individual in the cases was greater than that of the controls (ratio > 1), this finding did not reach a statistically significant association with case/control status for any of the minimum ROH length categories (Table 2).

Despite the average total length of ROH showing a general trend for increasing length with decreasing gestational weeks, neither regression nor ANOVA demonstrated a statistically significant change that was gestational week dependent (data not shown). In addition, there was no statistically significant association between total length of ROH and the case /control status as defined above via logistic regression, Table 3. We used PLINK to compute genome wide homozygosity in order to determine if there was more than expected overall homozygosity and whether there was a difference between preterm and term deliveries. For both cases and controls there was very little excess in homozygosity and we found no statistical association between the statistic and case/control status (data not shown).

ROH Mapping and Analysis

We next carried out mapping and analysis for ROH blocks with a minimum segment length of 2000 kb to access whether any local genomic regions were predictive of case-control status. A schematic of our methods appears in Figure 1. We found 424 50 kb segments with significant differences in the abundance of overlapping ROH blocks between the cases and the controls. The results are shown as a -log P Manhattan plot in Figure 2 with a threshold for p<0.05 shown. These were distributed randomly across the genome, Figure 3. We found that 88% of these regions had an overabundance of overlapping ROH in preterm mothers vs term mothers, a result not consistent with chance alone (binomial test p-value < .0001). While the p values for the individual 50 kb regions were not corrected for multiple comparisons, the results point to regions of interest or overlapping known genes associated with preterm birth. We mapped these significant 50 kb windows to the list of all known genes and found 199 known genes and 43 other transcripts including open reading frames, microRNA and lncRNA overlapping these regions and this data is available as supplementary material (see Supplementary Table S3 online). A list of the coordinates of these significant 424 50kb segments and the coordinates of their overlapping ROH blocks is available as supplementary material (see Supplementary Table S4 online).

Preterm Birth Associated Gene Mapping

From our previous work we found 30 pathways associated with preterm birth with high confidence values (FDR, <0.05) that contained 329 genes(19). We analyzed this gene set for overlap with all ROH blocks greater than 2000kb. Since 7 of the genes were from chromosome X, we excluded these and analyzed the remaining 322 genes. We found 3 preterm birth associated genes for which the overlapping ROH blocks were significantly more abundant in mothers delivering preterm. These genes (CXCR4, MYLK, PAK1) are shown in Table 4. We also examined the overlap with a mammalian gene set previously shown to have an evolutionary link to preterm birth with the ROH segments(20). We found ROH blocks with significantly greater abundance in preterm mothers overlapping CXCR4, PPP3CB, C6orf57, DUSP13, and SLC25A45 (Table 5) (20).

Discussion

Autosomal recessive inheritance can be investigated through homozygosity mapping or examination of large genomic segments for ROH. Previous population-based studies of consanguinity have demonstrated that there may be an increased risk of autosomal recessive inheritance associated with preterm birth(16, 17, 21, 22). Fine mapping of ROH has been made possible by high density SNP genotyping arrays. Homozygosity mapping has allowed identification of important genomic regions and recent studies have shown increased runs of homozygosity in intellectual disabilities associated with simplex autism(23). Using this approach we carried out a secondary analysis of a large genome wide association study for preterm birth using genotype data from the GENEVA Project. We sought evidence for structural variation (CNVs) and long runs of homozygosity in association with preterm birth. We used a restricted definition for preterm birth (<34 weeks) in order to avoid spurious associations. While we found no significant burden of runs of homozygosity, we did identify genomic regions with significantly greater abundance of ROH blocks in women delivering

preterm, albeit we did not correct for multiple comparisons, which overlapped genes known to be involved in preterm birth (Figure 2)

Out of the 322 genes identified in dbPTB(19), we found 3 genes (CXCR4, PAK1, MYLK) that overlapped with ROH blocks which showed significant abundance in cases versus controls. CXCR4 encodes a CXC chemokine receptor specific for stromal cell-derived factor-1. CXCR4 has been shown to be up regulated in labor and is related to inflammatory response(24). SDF1 needs CXCR4 as a receptor and it is "probable that the activation of CXCR4 by SDF1 is one of the sources of maternal-fetal immune tolerance"(25). SDF1 is upregulated during cytotrophoblast fusion to the syncytiotrophoblast.(26) It also facilitates trophoblast invasion into endometrium and enhances VEGF expression, which is crucial for placental angiogenesis(25). NFKB, an important inflammatory signaling molecule in the placenta, also leads to activation of CXCR4. Chemokines such as CXCR4 regulate decidual leukocyte recruitment during labor(27, 28).

PAK1 encodes a family member of serine/threonine p21-activating kinases, known as PAK proteins. PAK1 is only present in pregnant myometrial tissue(29). PAKS have been shown to regulate uterine contractility and/or load bearing during pregnancy(29). PKN1, is an alias of PAK1; an increase of PKN1 is associated with an increase in GTP RHOA (regulator of actin-myosin in uterine smooth muscle cells) in spontaneous preterm labor(30). Increases in contractile activity in term and preterm labor may be due to increase in RHO activity or RHO related proteins(30). Up-Regulation of myometrial RHO effector proteins (PKN1 and DIAPH1) is associated with increased GTP-RHOA in spontaneous preterm labor (29, 30).

MYLK, a member of the immunoglobulin gene superfamily, is expressed in muscle that encodes myosin light chain kinase. MYLK is involved in generation of smooth muscle tissue. Interestingly, studies show down regulation of MYLK along with the disparate regulation of MYL9, suggesting a mechanistic role of S-nitrosylation in preterm labor (31). This has been substantiated in a review of the human uterine smooth muscle S-nitrosoproteome fingerprint in pregnancy, labor and preterm labor(31).

These results are interesting from an evolutionary point of view. Risk alleles for preterm birth that deleteriously affect reproductive fitness should be eliminated by purifying selection(32). This phenomenon occurs more rapidly with greater efficiency of elimination for dominant alleles, allowing recessive alleles to persist in the population. This effect is called directional dominance(33). Directional dominance has been investigated particularly for behavioral traits where such observations have been made for schizophrenia and other forms of intellectual disability(33). We saw significant overlap of ROH blocks with genes associated with a mammalian gene set previously shown to have an evolutionary link to preterm birth(20). Interestingly, CXCR4 was identified both by mapping the genes linked evolutionarily to preterm birth and the genes identified by pathway-based analysis of a preterm birth GWAS(19, 20). PPP3CB is another of the evolutionarily linked genes. PPP3CB (also known as calcineurin A beta) is involved in the calcium influx dependent dephosphorylation of MEF-2A (which is involved in muscle development, neuronal differentiation, cell growth control)(31). Increased calcineurin A mRNA levels were found during pregnancy and studies have found the presence and activation of calcineurin/NFAT

signaling pathway just before labor in mice(34). In addition to the overlap with dbPTB genes and mammalian genes shown to have an evolutionary link to preterm birth, we found 199 genes overlapping the 50 kb significant windows.

Our study has both strengths and limitations. We examined a large collection of patients. This was a homogeneous population with very little substructure or stratification. The patients were individually genotyped on a high density array. We used a strict definition for prematurity (<34 weeks gestation) to avoid association with the more recent increases in late preterm infants. Our data were corrected for hemizygosity. We saw no association of copy number variations with preterm birth in this population. The limitations of the study include that this was a solely Danish population and we did not have a replication set.

We conclude that while we found no significant burden of runs of homozygosity, we did identify genomic regions with significantly greater abundance of ROH blocks in women delivering preterm that overlapped genes known to be involved in preterm birth. These data will be useful for future analysis of variants identified by deep sequencing or other strategies. They will help to prioritize candidate genes for further testing and/or biological validation.

METHODS

Case - Control Descriptions and Definitions

The Danish National Birth Cohort (DNBC) followed over 100,000 pregnancies beginning in the first trimester and has extensive biological material and epidemiologic data on health outcomes in both mother and child. Informed consent was obtained from all mothers as part of the DNBC data collection (35). The Gene Environment Association Studies initiative (GENEVA), funded by the trans-NIH Genes, Environment, and Health Initiative (GEI), dbGaP Study Accession: phs000103.v1.p1. consists of data for approximately 4000 Danish women and children from the DNBC, including phenotype and genotype information from approximately 1,000 mother-child pairs with mostly spontaneous onset of preterm labor or preterm premature rupture of membranes (PPROM) and 1,000 term birth mother-child pairs. Global inclusion criteria were applied to cases and controls, consisting of singleton gestation, live birth, child free of congenital abnormalities, absence of maternal conditions known to be associated with preterm delivery or often requiring early delivery of the baby (placenta previa, placental abruption, polyhydramnios, isoimmunization, placental insufficiency, pre-eclampsia/eclampsia). In addition, for inclusion in the GENEVA study, the child's parents and all four grandparents had to have been born in Denmark (except in 24 cases with one or two grandparents from other Nordic countries). For our interests, we limited our study to the genotype data of mothers that fell into two phenotypic groups: Controls includes mothers that went in labor at 38 weeks of gestation and later and Cases includes mothers that went into labor at 34 weeks of gestation and earlier. This analysis was approved by Institutional Review Board at Women & Infants Hospital of Rhode Island.

Genotyping

Genome-wide SNP genotyping was performed using the Illumina Human 660 W-Quad_v1_A, (Illumina, San Diego, CA) (n=560,768 SNPs). According to the data set release as well as the quality control and quality assurance policies of the GENEVA consortium(36), genotypes were not reported for any SNP which had a call rate less than 85% or which had more than 1 replicate error as defined with the HapMap control samples. Identical-by-descent (IBD) coefficients were estimated using 107,014 au-tosomal SNPs with missing call rate < 5% and no closer than 15 kb apart(36).

Population Structure and Relatedness

Population structure was examined by principal components analysis (PCA), as described by Patterson et al.(37), using independent, autosomal SNPs with missing call rates < 5% and MAF > 5%. To select independent SNPs, we utilized PLINK's LD pruning function(38). For the first round of short-range LD pruning we used a fifty SNP window with a shift of five SNPs, and pairwise genetic correlation with a threshold of 0.2. In a second round to remove long-range LD, we took the average number of SNPs over 5Mb from the output of the first round as our window (s = 131), again with an iteration of five SNPs and a threshold of 0.2. The resulting 69,877 SNPs were used to generate the principal components.

The relatedness between each pair of participants was evaluated by estimation of three coefficients corresponding to the probability that two (k2), one (k1) or zero (k0) pairs of alleles are identical-by-descent (IBD). IBD coefficients were estimated using 107,014 autosomal SNPs with missing call rate < 5% and no closer than 15 kb apart. All participants were analyzed together and all pairs of participants with a kinship coefficients > 1/64 were recorded.

Runs of Homozygosity

We used the Runs of Homozygosity (ROH) program in PLINK version 1.07 to identify autosomal ROH(38). The PLINK ROH algorithm is especially well-suited for SNP genotyping data analysis(39). Based on comparative studies among the ROH methods, PLINK was judged effective not only for SNP genotyping arrays but also on whole exome sequencing based data for long (1500 kb) ROH. We analyzed a range of minimum final segment lengths of ROH blocks including 1000 kb, 2000 kb and 3000 kb. Data from Chromosomes X and Y were excluded. We allowed for one heterozygous and five missing calls per window in order to prevent underestimating ROH as a result of an occasional genotyping error or missing genotype. In addition, the minimum density accepted was 50 (1 SNP per 50 kb) and the largest gap accepted was 1000 kb. In order to identify ROH that may be common in the general population, we used publicly available HapMap individual female genotypes from 30 CEU family trios (CEPH Utah residents with ancestry from Northern and Western Europe)(40). We used the genotype data from females only and ran ROH analysis to identify the possible common regions of ROH. We did not exclude any of these HapMap overlapping regions from our data, but instead we report those regions that overlap the previously identified genes associated with preterm birth from this Danish data set.

CNV Analysis

Since hemizygous deletions can mistakenly be designated as ROH, we analyzed copy number variations (CNV) to identify CNV regions that overlap with ROH blocks. We used Illumina's Bead Studio Program and PennCNV for CNV analysis. The results from the two different programs were largely similar. We used custom Perl scripts to identify the overlapping regions along the genome. For association testing we used PLINK to perform permutation-based test of association on segmental CNV data for cases and controls using the PennCNV variants.

Homozygosity Burden Analysis

The burden of autosomal homozygosity was analyzed by calculating the proportion of individuals with at least one ROH, the total number of ROH per individual, and the total length of ROH per individual for a given minimum final segment length. To determine the presence of statistically significant burden we used logistic regression with preterm/term status as the dependent variable for each of these predictors including maternal age at delivery as a covariate. Univariate analysis of the genotype data has demonstrated that the mother's age at delivery did not have any apparent effect and thus was left out of subsequent analysis. There was no significant difference in infants' gender and minimal difference in BMI as reported in GENEVA quality control document.

ROH Mapping and Analysis

We mapped and analyzed the abundance of ROH both genome-wide and by comparison to specific gene sets. First, we divided each chromosome (chr 1-22) into 50 kb windows and mapped the ROH blocks that overlapped with these windows. Any overlap was included and an ROH could be counted as overlapping multiple 50 kb windows. We applied Fisher exact test for each 50 kb window to compare the abundance of overlapping ROH in cases vs. controls. If a section carried at least 2 ROH blocks from cases or controls we included the section in the calculations. We used UCSC Table Browser to map significant 50 kb windows to all genes which have any overlap.

We next analyzed two gene sets associated with prematurity for their overlap with our identified ROH(19, 20). For a given set of genes, the length of each gene from build Hg18 plus 5 kb upstream and 5 kb downstream extensions was used to map overlap between the ROH blocks and the genes of interest. To be consistent in our approach, we again applied the rule: if at least two ROH blocks from cases or controls overlapped the gene we applied Fisher exact test to determine the significance of differential abundance in overlap between cases and controls for the gene.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

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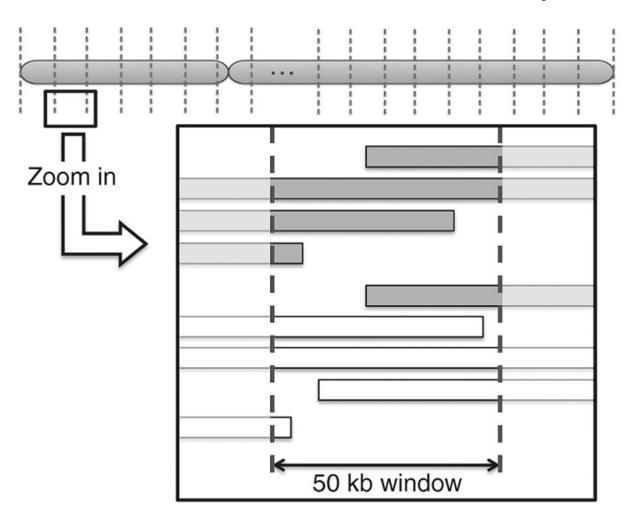


Figure 1. Schematic of ROH Analysis

Each chromosome (chr 1-22) was split into 50 kb wide windows. The number of minimum final segments of 2000 kb ROH blocks from cases and controls that fell into these sectors was analyzed for differential abundance by Fisher's exact test. Cases were represented in gray colored bars and controls were represented in white colored bars.

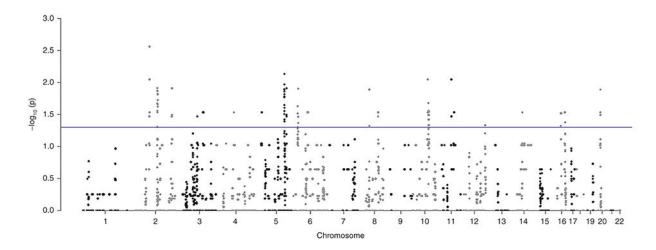


Figure 2. Manhattan Plot of p-values for genome wide 50 kb windows

The vertical axis shows the $-\log p$ value for comparison by Fisher's exact test of the number of ROH blocks nested within between preterm birth and control patients. Each dot in the graph represents the midpoint of 50 kb window for cases (34 weeks and less) and controls (38 weeks and above). The horizontal line shows the threshold for comparison of the abundance of blocks where the $-\log p$ is for p <0.05, above which difference in number of ROH blocks is significant.

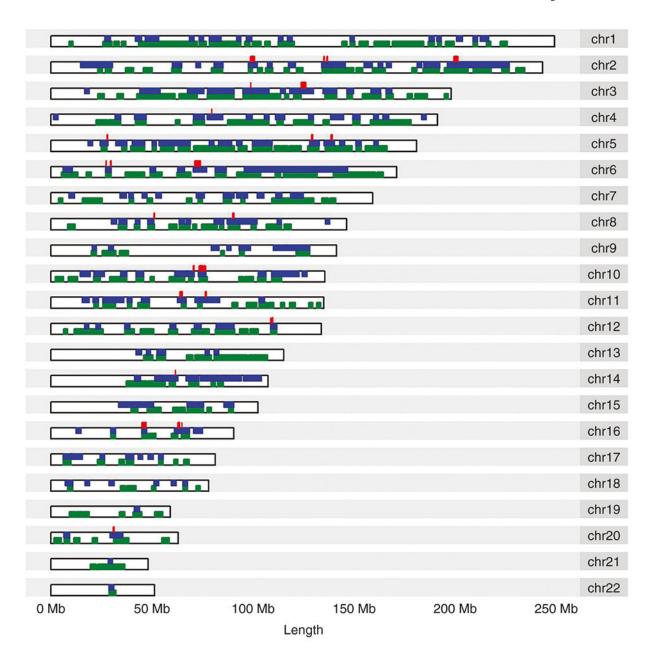


Figure 3. Chromosomal distribution of ROH and 50 kb significant windows
Green colored regions represent presence of a 2000 kb or greater ROH in at least one TERM patient. Blue colored regions represent presence of a 2000 kb or greater ROH in at least one PRETERM patient. 424 50 kb significant windows are marked in red.

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Table 1

Minimum segment length	Proportion of cases with ROH	inimum segment length Proportion of cases with ROH Proportion of controls with ROH Ratio Logistic regression coefficient P-value	Ratio	Logistic regression coefficient	P-value
1,000 kb	1	1	1	I	I
2,000 kb	0.733	0.689	1.06	-0.2181	0.084
3,000 kb	0.182	0.16	1.14	-0.1609	0.279

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Table 2

Minimum segment length	Average number ROH in cases	inimum segment length Average number ROH in cases Average number ROH in controls Ratio Logistic regression coefficient P-value	Ratio	Logistic regression coefficient	P-value
1,000 kb	18.008	17.655 1.02	1.02	-0.015	0.192
2,000 kb	2.081	1.923	1.923 1.08	-0.036	0.254
3,000 kb	1.602	1.515	1.515 1.06	-0.017	0.747

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Table 3

Minimum segment length	Average total ROH in cases	linimum segment length Average total ROH in cases Average total ROH in controls Ratio Logistic regression coefficient P-value	Ratio	Logistic regression coefficient	P-value
1,000 kb	25,530,560	24,627,939 1.04	1.04	$-7.489{\times}10^{-9}$	0.151
2,000 kb	5,800,931	5,205,360 1.11	1.11	-6.456×10^{-9}	0.346
3,000 kb	7,452,289	6,634,542 1.12	1.12	-3.952×10^{-9}	0.632

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Table 4

Gene symbol Name	Name	HGNC ID	Chr	HGNCID Chr Abundance P-value	P-value
CXCR4	Chemokine (C-X-C motif) receptor 4	2561	2	> Case	0.007
PAKI	P21 protein (Cdc42/Rac)-activated kinase 1	8590	11	11 > Case	0.02
MYLK	Myosin light chain kinase	7590	3	> Case	0.02

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Table 5

	Gene symbol Name	Name	HGNC ID	Chr	HGNC ID Chr Abundance P-value	P-value
Gain of function	CXCR4	CXCR4 Chemokine (C-X-C motif) receptor 4	2561 2	2	> Case	0.007
	PPP3CB	Protein phosphatase 3, catalytic subunit, beta isozyme	9315	10	> Case	0.027
	C60rf57	Chromosome 6 open reading frame 57	20957	9	> Case	0.029
Loss of function	DUSP13	Dual specificity phosphatase 13	19681	10	> Case	0.032
	SLC25A45	Solute carrier family 25, member 45	27442	11	> Case	0.033