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Multiple, independent, common variants at RET, SEMA3 and NRG1 gut enhancers specify Hirschsprung disease risk in European ancestry subjects

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Abstract

Purpose: Hirschsprung disease (HSCR) is a developmental disorder of the enteric nervous system (ENS) characterized by congenital aganglionosis arising from coding variants in ENS genes causing partial or total loss-of-function. Low-penetrance, common, noncoding variants at RET, SEMA3 and NRG1 loci are also associated with HSCR, with small-to-moderate loss of gene expression mediated through sequence variants in *cis*-regulatory elements (CRE) as another causal mechanism. Since these latter variants are common, many individuals carry multiple risk variants. However, the extent and combinatorial effects of *all* putative CRE variants within and across these loci on HSCR is unknown.

Methods: Using 583 HSCR subjects, one of the largest samples of European ancestry studied, and genotyping 56 tag variants, we evaluated association of all common variants overlapping putative gut CREs and fine-mapped causal variants at RET, SEMA3 and NRG1.

Results: We demonstrate that 28 and 8 tag variants, several of which are genetically independent, overlap putative-enhancers at the RET and SEMA3 loci, respectively, as well as two fine-mapped tag variants at the NRG1 locus, are significantly associated with HSCR. Importantly, disease risk

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Conflicts of Interest

The authors have no conflicts of interest to report.

increases with increasing numbers of risk alleles from multiple variants within and across these loci, varying >25-fold across individuals.

Conclusion: This increasing allele number-dependent risk, we hypothesize, arises from HSCRrelevant ENS cells sensing the reduced gene expression at multiple ENS genes since their developmental effects are integrated through gene regulatory networks.

Keywords

Hirschsprung disease; enhancer; polymorphism; genetic risk

1. Introduction

Hirschsprung disease (HSCR) is a developmental disorder of the enteric nervous system (ENS), characterized by loss of enteric ganglia in the myenteric and submucosal plexuses of the gut [1]. Failure of the precursor enteric neural crest cells to proliferate, differentiate, migrate and/or colonize the gastrointestinal tract leads to HSCR, the most common (15 per 100,000 live births [2]) cause of functional intestinal obstruction in neonates and infants [3]. The ensuing aganglionosis is caudal to rostral and based on the extent of aganglionosis, patients can be classified into short-segment HSCR (S-HSCR; aganglionosis limited up to the upper sigmoid colon, ~80% of cases), long-segment HSCR (L-HSCR; aganglionosis up to splenic flexure and beyond, ~15% of cases) or total colonic aganglionosis (TCA; aganglionosis of the entire large intestine, ~5% of cases) forms [1]. Nearly 80% of cases have aganglionosis only; in the remainder, HSCR co-occurs with multiple congenital anomalies, specific syndromes and/or chromosomal variants [1,4]. The critical features of HSCR are its high heritability (>80%), complex inheritance pattern, sex bias (4:1 affected male: female) and high recurrence risk to siblings and other relatives, influenced by sex, segment length of aganglionosis, familiality and the presence of additional anomalies [2]. An unsolved problem is to explain these diverse phenotypic features and associations with a single developmental mechanism.

Genetic studies in HSCR have uncovered rare, high-penetrance, coding variants in 14 genes $[4–6]$ and common, low-penetrance, noncoding variants near $RET[6–8]$, $NRGI[9]$ and the *SEMA3* gene cluster [6]. The pathogenic coding alleles at these critical ENS genes are partial or total loss-of-function alleles in heterozygotes, and collectively lead to a population attributable risk (PAR) of $~18.2\%$ [10]. In contrast, the polymorphic noncoding risk variants at RET and SEMA3 lead to a much larger PAR of ~37.7% [10]. Although these noncoding variants at RET and SEMA3 individually confer low-to-moderate risk (odds ratio, OR: 1.6– 3.9), collectively, they contribute to a 30-fold risk variation across individuals depending on risk allele dosage [11]. While reduced RET expression mediated by noncoding regulatory variants is the mechanism at RET, the causal molecular basis of associations at NRG1 and SEMA3 remain unexplored. Functional studies at the RET locus, have proven that three risk variants lead to reduced RET expression by disrupting SOX10, GATA2 and RARB binding at three functionally distinct, but synergistically acting, RET gut enhancers or *cis*regulatory elements (CREs) [8,12]. Although these three RET CRE variants are genetically independent (based on linkage disequilibrium (LD) in controls) they have synergistic effects on risk [11,12].

Given the widespread nature of multiple putative CREs for human genes, uncovered by the ENCODE [13] and NIH RoadMap Epigenomics [14] Projects, and supported by studies in model systems [15], along with extensive common noncoding variation in humans [16,17], it is a priori likely that multiple CRE variants at any given disease-associated locus can affect gene expression and thereby modulate disease risk, as we have shown for $RET[12]$. When multiple CRE variants affect a target gene's expression in *cis* there may be 'haplotype effects' independent of the LD between risk variants [12]. Thus, if larger numbers of variants lead to more extreme gene expression changes, and concomitantly higher risk, then affected individuals must be enriched for multiple risk variants. Indeed, there is considerable evidence, from prior segregation [2], linkage [18,19], and association [11] studies, that multiple risk variants define HSCR in individuals. In other words, mechanistically, association loci are likely to have multiple causal CRE variants, within and between loci, modulating gene expression and disease risk [20].

In this study, we have attempted to identify multiple causal CRE variants by evaluating association of all common noncoding variants overlapping HSCR-relevant putative enhancers at the RET and SEMA3 loci with HSCR risk, in a large sample of European ancestry cases. We also include all fine-mapped variants at the NRG1 locus [21], associated with HSCR risk in Asian ancestry subjects, to reassess whether this association also exists in this larger European ancestry sample. We indeed demonstrate that multiple, statistically independent, putative enhancer-overlapping variants at the RET and SEMA3 loci, and finemapped variants at the NRG1 locus, are significantly associated with HSCR risk. We also demonstrate that this risk increases as a logistic function with increasing risk allele dosage from multiple variants within and across these loci, suggesting that enteric neurons can quantitatively assess gene expression status across these, and presumably other, ENS genes. We hypothesize that this integration occurs through gene regulatory networks (GRNs).

2. Materials and Methods

2.1 Patient samples:

We analyzed HSCR patients and their family members ascertained from two sources: ongoing family studies from this (Chakravarti) laboratory (AC-HSCR) and by the Hirschsprung Disease Research Collaborative (HDRC; Appendix S1). The AC-HSCR collection includes participants recruited from referrals by practicing physicians, genetic counselors, family members and from self-referral through a study website and support group postings ($n=461$ families). The HDRC is a multi-institutional study that recruits participants from newly diagnosed cases from pediatric surgery and clinical centers at HDRC institutions $(n=276 \text{ families})$. The majority of AC-HSCR and HDRC participants are of self-described European ancestry.

2.2 Ethics statement:

All individuals were ascertained with written informed consent approved by the Institutional Review Board of the enrolling institution. This study is based on samples approved by the Institutional Review Board of the Johns Hopkins School of Medicine (March 31, 2018).

2.3 Variant selection:

For the RET (Figure S1) and SEMA3 (Figure S2) loci, we defined the target regions based on HSCR genome-wide association study (GWAS) signals [6] and flanking recombination hotspots from HapMap [16]. Within these regions we selected common (minor allele frequency (MAF) >10%) variants observed in 1000 Genomes [17] non-Finnish European (NFE) ancestry $(n=404)$ controls that overlapped putative gut enhancer marks from public epigenomic databases [13,14], as described earlier [12] (Dataset S1). Due to lack of prior evidence for association at *NRG1* locus in European ancestry HSCR subjects, we did not perform the extensive association screen but selected all 17 fine-mapped variants reported to be associated with HSCR in Asian ancestry subjects (Dataset S1) [21]. All variants selected at a locus fall within the same topologically-associated domain as the underlying risk gene (RET, SEMA3D and NRG1) [22]. Using 1000 Genomes [17] NFE ancestry genotype data and the software Tagger [23], 42, 11 and 7 tag variants at the RET, SEMA3 and NRG1 loci, respectively were selected with a LD r^2 cutoff of 0.5. We added the top two GWAS hits at SEMA3 locus (rs11766001 and rs12707682) [6] for a total of 62 variants that underwent multiplex genotyping assay design (Dataset S1).

2.4 Multiplex genotyping assay:

All steps of genotyping, including the design of custom multiplexed assays using iPLEX chemistry, were performed following the manufacturer's recommendations (Agena Bioscience, San Diego, CA, USA). Variants that failed in silico assay design or failed in genotyping of reference DNA samples were replaced with a best proxy variant from the filtered sets of variants when available; if not, we used the best replacement from the larger sets of common variants (based on LD) to generate two multiplexed assay pools for 61 tag variants (Dataset S1).

2.5 Variant genotyping quality control (QC):

Genotyping of 2,205 unique DNA samples was carried out for each multiplexed pool. After sample and variant QC (Supplementary Methods, Dataset S1, Table S1, S2), we combined variants across the two pools (31 and 25 variants in pools #1 and #2, respectively) to generate genotype data for 56 variants in 1,959 samples from 724 families. Genotypes at each variant from all unrelated HSCR probands $(n=583)$ were tested for Hardy-Weinberg Equilibrium (HWE): none were significant $(P<1.0\times10^{-6})$ except rs2435357 ($P=2.69\times10^{-19}$; Table S3), which was expected owing to its high population-level association with HSCR [7,8]. None of the HWE tests using genotypes from 1000 Genomes [17] NFE ancestry subjects were statistically significant (Table S3).

2.6 Statistical genetic analyses:

For control allele frequencies, we used publicly available data from the Genome Aggregation Database (gnomAD) [24], based on whole-genome sequencing of NFE ancestry subjects. These allele frequencies, determined through a different technology than our genotyping, were highly concordant to our pseudo-control data from HSCR trios (Figure S3). All HSCR cases with complete information on sex, segment length of aganglionosis and familiality, were classified into 8 risk categories, scored by known HSCR risk factors

(score of 0 each for male, short segment and simplex versus a score of 1 each for female, long segment/TCA and multiplex), and eventually summed into risk scores between 0–3 (Table S4) [25]. Population-level disease penetrance/prevalence for each risk allele bin was estimated using Bayes' theorem with the observed background control frequency and a disease incidence of 15 per 100,000 live births [8]. Polygenic risk scores (PRS) were calculated as the sum of risk alleles at HSCR-associated variants weighted by log_{10} transformed OR from case-control tests (AC-HSCR and HDRC combined cases vs. NFE controls). Statistical significance thresholds for single variant-based tests were adjusted for multiple tests using the Bonferroni correction for 56 tests at $P \le 8.9 \times 10^{-4}$. All P-values reported are two-tailed.

3. Results

3.1 HSCR subtypes differ between cohorts

We used two HSCR sample cohorts in this study: AC-HSCR and HDRC. AC-HSCR, collected through this laboratory, is likely to over-sample high-risk families while HDRC collected at clinical centers at first surgical encounter is likely to be representative of HSCR in the general population (Appendix S1). To assess if ascertainment differences can influence our analysis (see below), we compared the proportions of HSCR subtypes by sex, segment length (S-HSCR versus L-HSCR/TCA) and familiality in probands across the two cohorts (Table 1). Ignoring samples with unknown status, there was no significant difference between the cohorts for sex (71% vs. 76% males in AC-HSCR and HDRC, respectively; P=0.22). However, HDRC was significantly enriched for S-HSCR and simplex cases as compared to AC-HSCR (73% vs. 60% S-HSCR, P=0.007; 85% vs. 66% simplex, P=2.1×10⁻⁶ in HDRC and AC-HSCR, respectively). This confirms the expectation that AC-HSCR was enriched for high-risk families, i.e. those with L-HSCR/TCA and with affected relatives. These data also suggest that, in unselected HSCR cases the frequency of males, L-HSCR/TCA and positive family history are 76%, 27% and 15%, respectively; in AC-HSCR these are 71%, 40% and 34%, respectively, and except for sex are considerably higher. We also classified each proband into 8 risk categories based on known HSCR risk factors [25] and four risk classes (scores 0–3; Table S4). Comparisons of risk scores 0, 1 and 2+3 (combined due to small sample sizes) between the two cohorts showed a highly significant difference ($P=3.3\times10^{-4}$), with groups 0 and 2+3 being more abundant in HDRC and AC-HSCR, respectively (Table S5). Thus, the two samples may reveal different genotype-phenotype correlations dependent on causal variants enriched in high- versus low-risk probands; also, the presence of causal association in one sample does not guarantee it in the other [8].

3.2 Widespread genetic association of RET, SEMA3 and NRG1 variants with HSCR

Previous studies have established that common noncoding variants at RET, SEMA3 and NRG1 are significantly associated with HSCR risk [6–9]. However, these studies were from gene discovery laboratories who may have over-sampled probands from high-risk families or suffered from the 'winners curse'. This is relevant because our first discovered genetic association had differing phenotypic features depending on HSCR risk factors [7,8,11]. Our goal here, for the RET and SEMA3 loci, was therefore to assess the numbers and

degree of genetic associations at functionally relevant CREs, in HSCR cases including randomly ascertained new samples (HDRC), and to test for the presence of independent signals within each locus. We also wished to reassess the HSCR association [9] at $NRG1$ in European ancestry subjects. We genotyped a total of 2,205 subjects across 737 families from AC-HSCR and HDRC for 61 tag variants (MAF>10%), which after QC (Dataset S1, Tables S1–S3) reduced to genotype data from 56 tag variants in 1,959 subjects (701 affected) from 724 families (583 probands). From these, data from 583 independent HSCR cases were used for association analyses (Table 1). Of the 375 AC-HSCR independent cases, 320 have been evaluated earlier for association with limited set of risk variants at RET, SEMA3 and NRG1, and are used here for direct comparisons to the HDRC samples [11,12].

We used population-based case-control analysis and compared allele frequencies between HSCR cases and NFE ancestry controls (Table 2). At RET, 27 of 37 tag variants showed significant association with HSCR, including three previously published independent associations at rs2435357 (60% vs. 27%, OR=4.0, P=1.1×10−123) [7,8,11], rs2506030 (56% vs. 39% at rs788261, OR=2.0, $P=2.8\times10^{-30}$, $r^2=0.98$ with 2506030) [6,11] and rs7069590 (84% vs. 77%, OR=1.5, $P=1.8\times10^{-7}$) [12]. Among all *RET* variants evaluated, the most significant association remained at rs2435357. The ORs observed here at rs2435357, rs2506030 and rs7069590 are very similar to that previously reported in European ancestry HSCR subjects, in overlapping samples from AC-HSCR and by others (ORs range: 3.9–6.7, 1.8–2.3 and 1.5–1.7 for rs2435357, rs2506030 and rs7069590, respectively; Table S6). At SEMA3, 8 of 12 tag variants showed significant association with HSCR, including two previously published GWAS hits at rs12707682 (31% vs. 23%, OR=1.6, $P=2.6\times10^{-11}$) and rs11766001 (19% vs. 14%, OR=1.5, $P=1.3\times10^{-6}$) [6,11]. The ORs observed here at rs12707682 and rs11766001 are also very similar to that reported in European ancestry HSCR subjects, in overlapping samples from AC-HSCR and by others (ORs range: 1.3–1.8 and 1.4–2.3 for rs12707682 and rs11766001, respectively; Table S6). The most significant association at SEMA3 was, however, at rs1228877 (45% vs. 33%, OR=1.6, $P=2.6\times10^{-16}$). At NRG1 locus, 2 variants showed significant association with HSCR, including one of the Asian ancestry HSCR GWAS hits at rs16879552 (5% vs. 3%, OR=1.9, P=2.7×10−6) [9], a new result in European ancestry HSCR subjects [6,11,26]. This new observation, previously not significant ($P=0.4$) in overlapping samples from AC-HSCR [11] (Table S6), is driven by the HDRC samples (see below). The most significant association at NRG1 was, however, at rs16879576 (5% vs. 2%, OR=2.3, $P=7.2\times10^{-9}$), also driven by the HDRC samples (see below). We also calculated parent-to-child transmission rates for all tag variants in 314 HSCR trios and observed over-transmission of risk alleles at 31 variants ($P<0.05$) (Table S7).

We calculated pairwise LD (r^2) between all tag variants at each of the three loci in HSCR cases, as well as the 1000 Genomes [17] NFE controls, to assess the underlying LD structure (Figures S4–S6) and identify associations at independent genetic variants. Based on these results, several new and independent association signals at RET (Figure S4) and SEMA3 (Figure S5) are evident. However, the two NRG1 variants rs16879552 and rs16879576 are in moderate LD $(r^2=0.53/0.73$ in controls/cases) (Figure S6) and may represent a singular causal association. Unlike the RET locus, $SEMA3$ and $NRGI$ loci have been largely associated with HSCR risk in European versus Asian ancestry subjects, respectively. Having

detected new significant associations at NRG1 variants in European ancestry subjects here, we assessed if allele frequency differences between the two ancestries, in addition to ascertainment differences (see below), could explain these observed population-specific associations. Indeed, in general, allele frequencies for tag variants, including the GWAS hits at *SEMA3* and *NRG1* loci are quite different between NFE and East Asian gnomAD controls [24] (Figure S7).

Given the ascertainment differences in AC-HSCR versus HDRC, we performed association analysis for all tag variants in them separately (Table S8). At SEMA3, the risk allele at rs12707682 was at higher frequency and led to an increased OR in HDRC (23%/29%/36% allele frequency in controls/AC-HSCR/HDRC with significant ORs of 1.4 (95% CI: 1.17– 1.63) vs. 1.9 (95% CI: 1.55–2.33) in AC-HSCR and HDRC, respectively). However, within the same locus, significant risks at rs11766001 and rs1228937 were only evident in AC-HSCR (14%/21%/16% and 20%/28%/23% allele frequency in controls/AC-HSCR/HDRC with ORs of 1.7 (95% CI: 1.38–2.00) vs. 1.1 (95% CI: 0.87–1.50) and 1.6 (95% CI: 1.32–1.84) vs. 1.2 (95% CI: 0.98–1.55) in AC-HSCR and HDRC for rs11766001 and rs1228937, respectively). The confidence intervals for the above three SEMA3 variants overlap and so the estimates are not significantly different between the two cohorts. At NRG1, significant risks at rs16879552 and rs16879576 were only observed in HDRC (3%/4%/8% and 2%/3%/7% allele frequency in controls/AC-HSCR/HDRC with ORs of 1.3 (95% CI: 0.85–1.88) vs. 3.1 (95% CI: 2.18–4.48) and 1.7 (95% CI: 1.09–2.51) vs. 3.6 (95% CI: 2.44–5.36) in AC-HSCR and HDRC for rs16879552 and rs16879576, respectively), in accord with results from our previous study of overlapping AC-HSCR cases [11] (Table S6). At the RET locus, there were several variants with differential effects on HSCR risk between AC-HSCR and HDRC, including five with significant risk in AC-HSCR cases only (rs3758514, rs788243, rs7069590, rs12764797 and rs2472737), five with significant risk in HDRC cases only (rs3004255, rs4949062, rs3026693, rs1800860 and rs10793423), and at least five with substantially increased risk in HDRC as compared to AC-HSCR (rs2435357, rs2506024, rs2505515, rs2505513 and rs7074964) (Table S8). Overall, these results demonstrate that generally we detect a greater magnitude of associations in HDRC as compared to AC-HSCR, i.e., in samples enriched for simplex S-HSCR cases. For further analyses, any variant with statistically significant association in AC-HSCR, HDRC or AC-HSCR and HDRC combined cases was considered to be associated with HSCR risk.

We expect common variants to contribute differential risk across HSCR subtypes and, depending on the proportions of HSCR subtypes across cohorts, to overall risk. To evaluate this expectation we classified all HSCR subjects (AC-HSCR and HDRC combined) into four risk score groups (0, 1, 2, and 3) based on HSCR subtypes, as outlined above (Table S4) and performed case-control association tests for all variants across the three loci. Figures S8, S9 and S10 show the odds ratios and confidence intervals for each variant at SEMA3, NRG1 and RET locus, respectively, by risk score 0, 1 and 2+3 (combined due to small sample size). As expected, common variants associated with overall HSCR risk contribute the largest risk to group 0, followed by group 1, with largely no significant risk to the $2+3$ group. Unsurprisingly, the variants with no significant effects shows no such trend.

3.3 Risk of HSCR is additive across variants and genes

Having observed multiple single variant associations at RET (28 variants; Tables 2, S8; Figure S11), SEMA3 (8 variants; Tables 2, S8; Figure S12) and NRG1 (2 variants; Tables 2, S8), we next assessed their cumulative effects, within and across loci, on HSCR risk (Tables 3, S9–S10; Figure 1, S13). Here, we analyzed the two collections as one sample. For within locus analysis, given the rarity of risk alleles at the significant *NRG1* variants rs16879552 and rs16879576, we did not assess their effect. Risk alleles across all significant variants within RET and SEMA3 loci were counted in each individual with neighboring classes combined to generate five non-overlapping bins to reduce numbers of tests and to have similar sample sizes across bins in controls (Tables S9, S10). Similarly, for analysis across loci, risk alleles at all significant variants were counted in each individual and combined to generate five non-overlapping bins (Table 3).

At RET, the total risk allele dosage was significantly associated with HSCR risk (Table S9; χ^2 =134.7, P=3.9×10⁻²⁸; Figure S13) with the first three bins showing significant protection (OR=0.3/0.4/0.5 and P=3.2×10⁻⁸/2.8×10⁻⁶/6.7×10⁻⁵, for 17–23, 24–26 and 27–29 risk alleles, respectively) and the highest risk alleles bin having significantly increased disease risk (OR=4.9, $P=2.8\times10^{-26}$ for 35–52 risk alleles). The logistic function describing the relationship between HSCR risk and RET risk allele counts explains 80% of risk variation (Figure S13). We estimated the population-level penetrance (population probability of being affected given a genotype), which ranged from 4.9 cases to 40.0 cases for the lowest (17– 23 risk alleles) and the highest (35–52 risk alleles) bins, respectively (Table S9). These values translate to a population incidence difference ranging from \sim 1/20,000 to \sim 1/2,500 live births. At *SEMA3*, although there was a trend for an analogous association with risk allele dosage (Table S10; χ^2 =11.9, P=0.018; Figure S13), only two risk alleles bins (1–3) risk alleles, OR=0.7, $P=0.014$; 8–9 risk alleles, OR=1.5, $P=0.019$) showed a trend towards statistical significance. The logistic function describing the relationship between HSCR risk and SEMA3 risk allele counts explains 78% of risk variation (Figure S13). Counting risk alleles across the three loci led to an expected association between total risk allele dosage and HSCR risk (Table 3; χ^2 =156.4, P=8.5×10⁻³³; Figure 1), which, however, was more significant when compared to risk from the RET locus alone (Table S9; χ^2 =134.7, P=3.9×10⁻²⁸; Figure S13). The risk for HSCR from these variants increased from 0.2 to 5.4, a 26-fold change, as the bin size increased from 19–29 to 43–66 risk alleles. Also, the logistic function describing the relationship between HSCR risk and risk allele counts across the three loci explains 92% of risk variation, more than explained from RET alone (Figures 1, S13). The estimated population-level penetrance ranged from 3.7 cases to 45.4 cases for the lowest (19–29 risk alleles) to the highest (43–66 risk alleles) bins, respectively (Table 3). These values translate to a population incidence difference ranging from ~1/27,000 to \sim 1/2,200 live births. Collectively, these results show a clear additive effect of RET risk variants on HSCR risk and provide evidence for minor additional effects from SEMA3 and NRG1 risk variants.

Next, we wanted to assess how the combined (polygenic) risk from all the 38 significantly associated variants (8 SEMA3, 2 NRG1 and 28 RET variants) varied across HSCR subjects. PRS for each HSCR subject was calculated and varied from 5.12 to 16.04 (median=10.66)

and from 6.47 to 16.46 (median=11.28) in AC-HSCR and HDRC cohorts, respectively. Although the PRS variation followed normal distributions in both cohorts with mean values of 10.60 and 11.19 for AC-HSCR and HDRC, respectively, there was indeed a small yet significant difference in the two sample means ($z=2.63$; $P=0.009$; Figure 2). We also calculated PRS after LD r^2 -based pruning of significantly associated variants at each locus to remove moderately correlated variants $(r^2>0.2)$ leaving 3, 1 and 11 variants at *SEMA3*, NRG1 and RET locus, respectively. PRS varied from 1.63 to 6.65 (median=4.55) and from 1.87 to 7.06 (median=4.94) for in AC-HSCR and HDRC cohorts, respectively. As observed above with the PRS distribution based on all associated variants, the two cohorts show small but significant difference in polygenic risk distribution with relatively higher risk in the HRDC cohort (mean of 4.48 and 4.77 for AC-HSCR and HDRC, respectively; $z=2.71$; $P=0.007$; Figure S14). We expect the difference in composition of HSCR subtypes between the two cohorts (Table 1) to be the major factor leading to demonstrable variation in polygenic risk.

4. Discussion

Evidence in the literature [6–9,11,26,27], and the far more extensive data reported here (Table 2, Table S8), more than adequately confirms the role of multiple CRE polymorphisms at RET, SEMA3 and NRG1 on HSCR susceptibility in European ancestry subjects. In comparison with a *trans*-ethnic meta-analysis of GWAS for HSCR [28], the top RET variant here, rs2435357 is in high LD with the meta-analysis lead variant rs2505998 (r^2 =0.97 in Non-Finnish 1000 Genome European ancestry subjects), thus likely representing the same association signal. At NRG1, the top two associated variants here, rs16879552 and rs16879576 are not in LD with rs7005606, the meta-analysis lead variant (r^2 of 0.024 and 0.011 in Non-Finnish 1000 Genome European ancestry subjects, respectively), thus likely representing a different association signal. In addition, rs7835688, which is in complete LD with rs7005606 $(r^2=1$ in Non-Finnish 1000 Genome European ancestry subjects), is not associated with HSCR in our study. At SEMA3, the lead variant rs80227144 from the trans-ethnic meta-analysis is in weak LD with the best tagging variant here, rs11766001 $(r^2=0.23$ in Non-Finnish 1000 Genome European ancestry subjects) that is associated with HSCR in our study. The top two associated variants here, rs1228877 and rs12707682, are in very low LD with $rs80227144$ (r^2 0.072 and 0.149 in Non-Finnish 1000 Genome European ancestry subjects, respectively), thus likely representing a different association signal.

Whether these specific variants disrupt enhancer function to dysregulate these genes remain to be experimentally demonstrated, as we have previously shown for rs2435357, rs2506030, and rs7069590 at $RET[12]$. However, the most parsimonious hypothesis is that these gut enhancer variants reduce gene expression at RET, SEMA3 and NRG1 to increase the risk of aganglionosis, with risk increasing as a logistic function of the numbers of risk alleles (Figure 1). A first question is whether marked expression reduction at RET alone leads to HSCR given its large impact relative to $SEMAS$ and $NRGI$? This is unlikely because disease also occurs in non-risk allele homozygotes at RET, albeit at a lower rate. We have recently demonstrated [10], following years of indirect evidence from segregation [2], linkage [18,19] and association studies [11] that individual patients harbor multiple rare risk coding variants across different genes; here, we show the same feature with common CRE

risk variants. This prompts the second question: how does an enteric neuron 'count' risk variants, or recognize expression change, across diverse genes to reflect its potential for aganglionosis?

The mechanical and technical aspects of mapping variants to disease, long recognized through GWAS of complex traits, does not, however, clarify the underlying 'counting' mechanism even when we know that disease results from gene expression changes. For understanding disease pathophysiology, we need to understand the degree of expression change required for aganglionosis to result. Clearly, single variants, the units of mapping, show disease association. However, this does not imply that the variant alone leads to disease, indeed, it likely does not, because its genetic effect is usually small. Further, even 50% gene expression reduction at RET, in heterozygotes for null coding alleles, does not necessarily lead to a clinical phenotype [8]. Thus, there must be >50% expression reduction at each gene before disease results, a threshold unlikely to be reached by single variants or even single genes and their enhancers. Consequently, multiple functional variants act together to have a sufficient effect on gene expression to reach this threshold; the more extreme the threshold the larger the genetic effect whether from one or many variants.

The genetic consequences of this hypothesis is that these risk variants are 'counted' by enteric neurons. One possibility is that multiple risk variants physically exist on a single chromosome in cis to exert a significant effect on gene expression, irrespective of whether these variants are in LD or not. Indeed, the causal genetic variants may be correlated, leading to the suspicion that they do not impart independent effects; it is precisely this dependence (accumulation) that leads to a 'jackpot' effect on gene expression by that haplotype. In other words, LD changes the frequency of this variant haplotype but not its impact on the phenotype.

How then are the effects counted across genes? Current data suggests that the RET, SEMA3 and NRG1 effects on HSCR are additive, not unlikely given their small effect sizes. As we have shown for $RET[12]$, developmental genes function within a GRN and the effects of a set of variants on one gene affects the expression of other network genes. Thus, combinations of alleles at these three genes may impart greater risk through affecting all genes in the network. For ENS development, the known GRN regulates both RET and EDNRB, so that all variants that reduce their expression levels significantly will lead to aganglionosis. Proving this hypothesis using human genetic data alone is going to be arduous because HSCR is a rare disorder and difficult to collect thousands of samples. In contrast, mouse models with enhancer variants at Ret, Sema3 and Nrg1 can be used to test the contention that even with variants that reduce gene expression at developmentally important disease genes, disease only results from specific combinations of genotypes across these loci depending on the architecture of the GRN and three-dimensional chromatin contacts. Such a hypothesis may be sufficient to explain the need for multiple genes in a disease, multiple variants in these genes, small genetic effects of disease-associated alleles, the familiality of the disorder and its reduced penetrance.

This work also shows that the nature of the patients ascertained in a study can significantly impact the variants and genes identified. Past studies had a greater focus on the analysis

of rare variants in coding genes, which were successful in multiplex, severe and syndromic families. However, identification of regulatory defects in HSCR have succeeded mostly in simplex, short segment non-syndromic cases. Since the latter are far more common than the former, exploring the regulatory hypothesis in HSCR through GWAS is likely to be more fruitful for explaining disease risk.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

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List of abbreviations

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Figure 1: Logistic function relating Hirschsprung disease risk with the total number of risk alleles at *RET***,** *SEMA3* **and** *NRG1* **loci.**

The logarithm of odds ratios (Y-axis) for binned risk allele counts (X-axis) based on significant variants across the three loci are shown along with the best linear fit (dashed line).

PRS comparison

Figure 2: Polygenic risk score distributions. Empirical frequency distributions of polygenic risk scores (PRS) in AC-HSCR (blue) and HDRC (red) HSCR subjects are plotted along with their corresponding fitted normal distributions.

Table 1:

Distribution of HSCR probands from the AC-HSCR and HDRC studies by sex, segment length of aganglionosis and familiality.

 α ² Comparison between males and females,

b short and long+TCA cases, and

 ϵ simplex and multiplex cases, respectively.

Table 2:

Case-control association tests for tag variants at SEMA3, NRG1 and RET in HSCR.

The tag variants at each locus are listed in genomic order. The frequency of the coded allele (higher frequency in HSCR) at each variant in cases and controls, odds ratio with 95% confidence interval (CI) and the statistical significance of association (P) are provided. Multiple test-corrected significant P-values are in bold.

Table 3:

Odds ratio and risk of HSCR as a function of the number of risk-increasing variants at RET, SEMA3 and NRG1.

Overall χ**2**=156.43; P=8.52×10−33

The numbers of cases and controls classified by the number of risk alleles at RET, SEMA3 and NRG1 significant variants, odds ratio with 95% confidence interval (CI), statistical significance of association (P) and estimated population-level penetrance (prevalence) are provided. Statistically significant values are in bold. The penetrance value shown is the expected number of cases in 100,000 live births of that class assuming a population average of 15 cases per 100,000 live births.