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To cite this version:

Raquel Fernández, Marta Bleda, Berta Luzón-Toro, Luz García-Alonso, Stacey Arnold, et al.. Pathways systematically associated to Hirschsprung’s disease.. Orphanet Journal of Rare Diseases, BioMed Central, 2013, 8 (1), pp.187. 10.1186/1750-1172-8-187. inserm-00921630

HAL Id: inserm-00921630
https://www.hal.inserm.fr/inserm-00921630
Submitted on 20 Dec 2013
Pathways systematically associated to Hirschsprung’s disease

Raquel M Fernández1,2, Marta Bleda2,3, Berta Luzón-Toro1,2, Luz García-Alonso3, Stacey Arnold4, Yunia Sribudiani5, Claude Besmond6, Francesca Lantieri7, Betty Doan6, Isabella Ceccherini7, Stanislas Lyonnet6, Robert MW Hofstra5, Aravinda Chakravarti4, Guillermo Antiñolo1,2, Joaquín Dopazo2,3,8* and Salud Borrego1,2*

Abstract

Despite it has been reported that several loci are involved in Hirschsprung’s disease, the molecular basis of the disease remains yet essentially unknown. The study of collective properties of modules of functionally-related genes provides an efficient and sensitive statistical framework that can overcome sample size limitations in the study of rare diseases. Here, we present the extension of a previous study of a Spanish series of HSCR trios to an international cohort of 162 HSCR trios to validate the generality of the underlying functional basis of the Hirschsprung’s disease mechanisms previously found. The Pathway-Based Analysis (PBA) confirms a strong association of gene ontology (GO) modules related to signal transduction and its regulation, enteric nervous system (ENS) formation and other processes related to the disease. In addition, network analysis recovers sub-networks significantly associated to the disease, which contain genes related to the same functionalities, thus providing an independent validation of these findings. The functional profiles of association obtained for patients populations from different countries were compared to each other. While gene associations were different at each series, the main functional associations were identical in all the five populations. These observations would also explain the reported low reproducibility of associations of individual disease genes across populations.

Background

Unlike a minority of Mendelian traits, most human diseases have complex, multifactorial inheritance where the causation resides in small allelic differences in many genes occurring in a complex manner. For these phenotypes, onset, penetrance, recurrence risk, etc., are features not dependent on one single gene, but are rather emergent properties of the ensemble of genotypes at many loci [1]. A representative example of this kind of trait is Hirschsprung’s disease (HSCR, OMIM 142623), a neurocristopathy characterized by the absence of intramural ganglion cells in the myenteric and submucosal plexuses along a variable portion of the distal intestine. Based on the length of the aganglionic region, the disorder is classified into short segment (S-HSCR: aganglionosis up to the upper sigmoid colon, 80% of cases), long-segment (L-HSCR: aganglionosis up to the splenic flexure and beyond, 17% of cases) and total colonic aganglionosis forms (TCA, 3% of cases). The most widely accepted etiopathogenetic hypothesis for HSCR is based on a defect of craniocaudal migration of neuroblasts originating from the neural crest that, under normal circumstances, reach the small intestine in the 7th week of gestation and the rectum in the 12th week [2,3]. HSCR constitutes a complex pathology with non-Mendelian inheritance, sex-dependent penetrance, variable expression and suggestive of the involvement of one or more gene(s) with low penetrance [2,3]. With a relative risk as high as 200, HSCR can be considered an excellent model to study common multifactorial diseases. The major HSCR predisposing event is the presence of a haplotype at the RET proto-oncogene [4,5] (OMIM 164761, 10q11) which comprises a SNP lying in an enhancer element of the intron 1 [2,3,6-8]. To date, several HSCR-associated regions, such as 10q11 [9-12], 13q22 [10], 9q31 [9,13], 3p21 [11,14], 19q12 [11], 16q23 [10],
Conventional gene-based association tests present obvious limitations, especially in the context of rare diseases, where the recruitment of large cohorts of patients is extremely difficult. However, Pathway-based analysis (PBA) strategies [18-20], which allow for the detection of modules of functionally-related genes associated to the disease, have already been successfully applied to the study of a number of diseases [17,21-23]. The recent description of HSCR-associated functional modules in Spanish population [17] constitutes an excellent example of how a PBA strategy can be successfully applied to define the molecular basis of the mechanism of the disease. Here we have extended such approach to different populations of the International Consortium for Hirschsprung disease (ICHSCR) which have provided samples from France, Italy, the Netherlands and the USA. The extended study allowed us to conclude that functional modules related to signal transduction and its regulation, neurogenesis and the Ras pathway are common to all the populations in spite of the fact that the most associated genes in each population were different. Moreover, Network analysis recovers sub-networks significantly associated to the disease and populated by genes with the same functionalities, thus confirming the findings of the PBA analysis by an independent methodology.

Methods
General analysis strategy
Based in the recent study which identified four new loci associated to HSCR [17] as a starting point, we aimed to validate whether the discoveries made were specific for the analysed population or, on the contrary, were common mechanisms shared by other populations. Figure 1 shows the strategy followed in this study: our initial hypothesis was that disease genes would probably be population specific whereas functional modules would be universal.

Samples and SNP genotyping
We have conducted a genome wide genotyping of a multinational cohort of 162 trios of sporadic short-segment Hirschprung's patients recruited in the context of the ongoing initiative of the ICHSCR. Genotyping was carried out using the Affymetrix 500 k chip (composed
of the 250 k Nsp and the 250 k Sty chips) (Table 1). Given that not all the trios were simultaneously genotyped with both chips yet (only 26 of them), we considered two independent measurements, one for each chip (Sty and Nsp, with 97 and 91 trios, respectively), of the samples analyzed. Quality controls were as follows: SNPs missing in more than 20% of the samples in the calling process, SNPs with MAF < 0.5%, with Mendelian errors or not in Hardy-Weinberg equilibrium (in unaffected samples; p-value < 10^{-5}) and samples with more than 5% of the SNPs missing were discarded. Finally, incomplete families (trios) were also discarded. Additional file 1: Table S1 summarizes the effects of the different steps of the quality control.

Pathway-based analysis
We conducted a Transmission Disequilibrium Test (TDT) association analysis as implemented in the PLINK [24] software for the different sets of trios analyzed (see Figure 1).

The SNPs were ranked according to their p-values obtained in the TDT test and then a PBA test [18], as implemented in the GESBAP [25] module of the Babelomics [26] software, was conducted. PBA seeks for gene sets (GO terms in this study) associated to low p-values. This association is found significant when a number of genes of the GO term, larger than expected by chance, simultaneously display low (although not necessarily significant) individual p-values [18]. Given that only genes can be related to GO terms, PBA tests use only SNPs mapping on genes, or in the close neighborhood, here defined as 500 bps up-and downstream of the gene limits. When multiple SNPs map onto the same gene, the SNP with lowest (most significant) p-value is retained. In this way a list of genes ranked by the best of the p-values of all the SNPs mapping onto them is constructed. Then, GO terms significantly overrepresented among the genes associated to low p-values are found upon the application of a logistic regression. GO terms are declared significantly associated to HSCR when a number of its genes, larger than expected purely by chance, display a low p-value (i.e., are on the top of the list ranked by significance). To control the number of false positives due to multiple-testing effects, only GO terms with a FDR-adjusted [27] p-value < 0.05 are declared significant. The adjustment process takes into account that four populations and two chips have been tested, so all the individual tests were considered here. This is a widely accepted method for correcting p-values that account for multiple testing by controlling the rate of false discoveries. By default the GESBAP software only analyses GO terms between levels 3 and 13 of the GO hierarchy and exclude GO terms with more than 600 genes or with less than 5 genes. In this way GO terms which are either too specific or too unspecific, and only contribute to the decrease of the statistical power of the test, are avoided.

Network analysis
Like in the PBA approach, a list of genes ranked from low to high p-values is explored in order to discover subnets with connectivity values higher than expected by chance. Briefly, the N (10 in this case) most significant genes are mapped onto the interactome and the minimum network connecting them is obtained. The connectivity of such sub-network is calculated as the average of all the individual connectivity values of all the connected genes. The connectivity parameter accounts for the number of partners of direct interaction that a particular node has. An empirical distribution of the random expectation of this parameter can be obtained by repeatedly sampling random sets of N genes from the complete genome and calculating the average connectivity of their corresponding minimum connecting trees. Thus, the real value obtained for the N most associated genes can be contrasted with respect to its random empirical expectation. If the connectivity is not significantly higher than its random expectation we repeat the procedure for the N + 1 most significant genes. The procedure is repeated until a sub-network of significant connectivity is found or a value of N too high is reached (200 in this case). This procedure [28] is a generalization of the network analysis methodology applied to the study of networks contained in gene expression signatures [29,30]. An implementation of the procedure can be found in the Babelomics package [26].

**Table 1 Available trios for any of the chips distributed among the five country populations analyzed**

| Chip     | France | Italy | USA | Netherlands | Total |
|----------|--------|-------|-----|-------------|-------|
| Nsp      | 15     | 20    | 25  | 37          | 97    |
| Sty      | 16     | 18    | 25  | 32          | 91    |
| Nsp + Sty| 5      | 1     | 9   | 11          | 26    |
| Total trios | 26    | 37    | 41  | 58          | 162   |

First and second rows refer to the trios genotyped with each specific chip (Nsp or Sty). Third file contains the trios simultaneously genotyped with both chips (Nsp + Sty). Forth row is the total number of trios used in the study for each population.

**ENCODe information for extra Genic SNPs**
Extra genic SNPs have been used to find extra support for the functionalities found as associated to HSCR. We have used the HaploReg [31] tool to retrieve the relevant information from the ENCODE project [32] corresponding to the chromosomal regions in which significant extra-genic SNPs map.
Results
Validation of the known HSCR-associated functionalities by PBA of the available chips from the different country populations

As described in the Methods section (see Figure 1), we used the GO terms already described as part of the disease mechanism in the Spanish population [17] as the initial hypothesis that should be validated using all the available information in the four populations of the consortium. Then, an independent PBA was carried out for each chip in each population. The p-values of the SNPs in each chip/population were obtained by means of a TDT test, as implemented in the PLINK package. Such p-values were introduced in the PBA section of the Babelomics package to complete the PBA test and obtain lists of GO terms significantly associated to each condition tested.

Results are summarized in Table 2. All the GO terms previously proposed as components of the disease mechanism were validated in the four populations in one or both chips, with the exception of enzyme linked receptor protein signaling pathway (GO:0007167), which seems to be an aspect of signaling peculiar of the Spanish population. In addition to the proposed GO terms other new terms have been included in Table 2 because of their consistent significance across the populations analyzed and also because of their relationship to ENS formation. However, it is known that gene-based p-values are biased towards longer genes and those within weak LD region. Such genes have higher probability of encompassing more independent SNPs that will increase the probability of displaying significant p-values just by chance [33]. Although this bias is less expectable in GO terms, because these are composed by numerous genes with no a priori bias towards any particular size, it is true that they exist GO terms with an unexpectedly high number of large genes, whose p-values can be biased [34]. In order to check potential biases due to abnormally number of large genes in GO terms we have plotted the distribution of gene sizes of each significant class in Table 2 against the background distribution of sizes and only three of them were slightly bigger that the background distribution (see Additional file 2: Figure S1). Actually, three terms show a median size, which is higher than the 3rd quartile of the distribution of median gene sizes of all the GO terms (see Additional file 3: Figure S2). These GOs, however, are biologically related to the disease and related to other GOs without this size bias.

Figure 2 graphically summarizes a comprehensive map of the biological functionalities associated to the disease. GOs potentially affected by gene size bias are represented in a different color. As previously described, GO modules related to signal transduction and its regulation that include the parent regulation of signal transduction

| GO ID                | Initial hypothesis                        | Italy        | France       | Netherlands | USA          |
|----------------------|------------------------------------------|--------------|--------------|-------------|--------------|
|                      |                                          | Nsp | Sty | Nsp | Sty | Nsp | Sty | Nsp | Sty |
| Signaling            | Regulation of small GTPase mediated signal transduction | Y   |   | Y   |   | Y   |   | Y   |   |
| GO:00051056          | Regulation of Ras protein signal transduction | Y   |   | Y*  |   | Y   |   | Y*  |   |
| GO:0007265           | Ras protein signal transduction           | Y   |   | Y   |   | Y   |   | Y   |   |
| GO:0007264           | Small GTPase mediated signal transduction | Y*  |   | Y*  |   | Y*  |   | Y*  |   |
| GO:0035023           | Regulation of Rho protein signal transduction | Y   |   | Y   |   | Y*  |   | Y   |   |
| GO:0009966           | Regulation of signal transduction          | Y*  |   | Y*  |   | Y*  |   | Y   |   |
| GO:0007167           | Enzyme linked receptor protein signaling pathway | Y   |   | Y   |   | Y   |   | Y   |   |
| GO:0007268           | Synaptic transmission                      | Y   |   | Y   |   | Y   |   | Y   |   |
| ENS formation        | Calcium ion transport                      | Y   |   | Y*  |   | Y*  |   | Y*  |   |
| GO:0006812           | Cation transport                           | Y*  |   | Y*  |   | Y*  |   | Y*  |   |
| GO:0016337           | Cell-cell adhesion                         | Y*  |   | Y   |   | Y   |   | Y   |   |
| GO:0016477           | Cell migration                             | Y   |   | Y   |   | Y   |   | Y   |   |
| GO:0007399           | Nervous system development                 | Y   |   | Y   |   | Y   |   | Y   |   |
| GO:0048666           | Neuron development                         | Y   |   | Y   |   | Y   |   | Y   |   |
| GO:0007409           | Axonogenesis                               | Y   |   | Y   |   | Y   |   | Y   |   |

The first column specifies the general biological process represented by the GO terms (columns 2 and 3) which were found significant in the Spanish population and constitute the initial hypothesis for the functional basis of the disease. From fourth column ahead a Y means that the GO term was significant in the corresponding population and chip (FDR-adjusted p-values < 0.05) using the PBA approach described in Methods. Y* means that although this particular term was not significant, another direct descendant or ancestor GO term in the hierarchy was significant. The last four files are GO terms (in italics) that were not initially significant in the Spanish population but were consistently significant through the populations analyzed here.
(GO:0009966) and the rest of descendants (regulation of small GTPase mediated signal transduction, GO:0051056; regulation of Ras protein signal transduction, GO:0046578; and regulation of Rho protein signal transduction, GO:0035023), potentially affected by gene size bias, and the regulatory terms small GTPase mediated signal transduction (GO:0007264) and Ras protein signal transduction (GO:0007265). However, enzyme linked receptor protein signaling pathway (GO:0007167) was only confirmed by the USA population, suggesting that it is not a general mechanism of the disease but it rather represents a population-specific peculiarity. Finally, synaptic transmission (GO:0007268) a process whose malfunction is licit to consider to have a role in the disease, has also been validated in all the studied populations. On the other hand, cation transport (GO:0006812) and, in particular, the descendant term calcium ion transport (GO:0006816) are known to be altered in the disease. Both terms have been validated in the four populations analyzed. Finally, another functionality affected in all the populations, related to the formation of functional enteric cells, is cell-cell adhesion (GO:0016337). Of particular interest is the discovery of a series of GO terms involved in the generation of neurons that take part of the ENS, such as nervous system development (GO:0007399) and the descendant terms neuron development (GO:0048812) and axonogenesis (GO:0007409). These terms did not reach a significant p-value in the Spanish populations [17] but, no doubt these functionalities can be considered as part of the underlying mechanism of the disease.

Comparison of the gene functionalities associated to HSCR observed between country sample populations

Apart from the validation of the core disease related functionalities, the independent study of the functional associations obtained by PBA of the 4 different country populations of the ICHSCR rendered a considerable number of GO terms.

Again, when individual country populations are independently analysed two key processes become apparent: signalling and neurogenesis. In fact, the GO terms synaptic transmission, regulation of small GTPase mediated signal transduction, Ras protein signal transduction and nervous system development were significant in all the 5 analysed populations. Moreover, it is clear from the figures depicting the GO structure (Additional file 4: Figure S3...
and Additional file 5: Figure S4) that a relevant number of GO subcategories not shared by all the country populations actually belong to three main branches: nervous system development, signal transduction and cell migration. Additional file 6: Table S2 and Additional file 7: Table S3 show a total of 49 and 29 GO modules, respectively, significantly associated to HSCR found in the analysis individualized by population in any of the two chips. Additional file 4: Figure S3 and Additional file 5: Figure S4 show how all these modules are interrelated among them. For example, terms like axonogenesis, neuron development, neuron differentiation, generation of neurons, are all descendants of neurogenesis, which is itself, a descendant of nervous system development. Cell-cell adhesion is also a parent term of many other GO terms related with the formation of functional enteric cells. And there is also the case of other processes whose malfunction is relevant in the disease, such as synaptic transmission, the transport of several substances (cation transport or phospholipid transport) or different signalling-related functions, including the well-known RAS/RHO intracellular signalling pathway.

In order to know if the GO module associations found in each population were consistently a consequence of the underlying associations of the same specific genes, we compared the individual gene association values across all country populations. A list with the SNPs mapping within or in the neighborhood of genes ranked by association to HSCR (that is, the p-value of the TDT) was generated. Figure 3 shows the genes common to the four Consortium populations analyzed, plus the Spanish population already studied [17]. Only AGAP3 and TUBA8 genes are common to the five populations when the 400 SNPs showing the highest association to HSCR are considered. However, when the GO terms corresponding to the genes are compared across populations, a remarkable coincidence of affected functions occurs, as reflected in Table 2. While in the case of genes we need to reach up to 200 SNPs to have more than one common gene, and up to 500 SNPs to have a third common gene (apart from AGAP3 and TUBA8), when GO terms corresponding to the genes that appear in the rank of the first 10 SNPs are compared many coincidences are found. If we expand the percentile, the coincidences in functionalities increase enormously. All the coincident functionalities are related again to signaling and ENS formation.

Therefore, our observations strongly suggest that common functions affected in different populations are not due to common genes but rather to different genes of common functionality, which are affected in different populations and cause similar phenotypic effects.

Network analysis
The lists of genes from both chips, ranked by association to HSCR (according to SNP p-values obtained from the corresponding TDTs), were scanned to find sub-networks
of interacting proteins significantly associated to the disease [28]. The application of this network analysis to the Nsp chip rendered a significant network (p-value = 0.027; see Methods) within the first 27 most associated genes, linking 19 of them. In the derivation of the network a maximum of one external node linking these proteins was allowed [28]. This significantly large sub-network (Figure 4) significantly associated to the disease documents the basis of the complexity of this disorder. Genes like GABBR2, GRIN2B and HTT associated to synaptic transmission (GO:0007268), and the last one also associated to central nervous system development (GO:0007417) and neuron development (GO:0048666) are among the 27 most associated genes, along with RET, also in the list. Interestingly, genes connecting them and included in the network significantly associated to the disease are: CRK, related with regulation of small GTPase mediated signal transduction (GO:0051056), regulation of Ras protein signal transduction (GO:0046578) and Ras protein signal transduction (GO:0007265); GRB2, related to Ras protein signal transduction (GO:0007265); ITGB1, related to cell-cell adhesion (GO:0016337), cell migration (GO:0016477), cell projection organization (GO:0030030) and neuron development (GO:0048666); PLCG1, related to cell migration (GO:0016477); RPS27AP16, related to synaptic transmission (GO:0007268), cell projection organization (GO:0030030) and neuron development (GO:0048666); SH3GL3 and TP53, related to central nervous system development (GO:0007417). Again, the main processes highlighted by the network analysis are signalling and neurogenesis. Moreover, the closeness in the interactome of the HSCR genes to the genes selected by the network analysis strongly supports the relationship of the sub-network to the disease. The application of this network analysis to the Sty chip results in a significant network (p-value = 0.04; see Methods) within the first 67 most associated genes. In the derivation of the network a maximum of one external node linking these proteins was allowed [28] (see Additional file 8: Figure S5). This larger network contains some of the genes already linked in the Nsp network plus some others and point exactly to the same affected functionalities.

Extra-genic SNPs associated to HSCR that support the functionalities found

Very recently the results of the ENCODE project have been published, allowing to assign functionalities to

![Figure 4](image.png)

**Figure 4** Significant sub-network of 53 genes, allowing for one intermediate gene, associated to HSCR (p-value = 0.027) in the Nsp chip. The network analysis [28] is carried out by using the functional analysis options (Set enrichment analysis/NetworkMiner) of Babelomics [26] on the list of genes ranked by the p-values obtained upon the application of a conventional TDT association test as implemented in PLINK [24].
regions in which some extra-genic SNPs associated to HSCR were found to map. Table 3 lists six extra-genic SNPs with a nominal p-value < 10^{-5} obtained for each chip analyzing all the trios together, for which some annotation has been found. Thus, rs2435367, one of the SNPs with a lowest p-value, is mapping in a DNAse hypersensitivity region, typically related to gene expression, close to the 5' of RET. The SNP rs2505526 (p-value = 8.353×10^{-7}), changes a transcription factor binding site (TFBS), Pax-4, close to the RASGEF1A gene, related to signal transduction (particularly the Ras pathway) and ENS formation. The SNP rs16838932 (p-value = 4.459×10^{-7}) maps in a ultraconserved [35] region and it is difficult to speculate any relation with the disease by itself or through the closest gene, SLC39A10. The SNP rs2659635 (p-value = 7.098×10^{-6}) maps in a TFBS, Foxj2, in an ultraconserved [35,36] genomic region. The SNP rs4570660 (p-value = 7.744×10^{-6}) maps in a DNAse hypersensitive region and, within it, in the binding site of the transcription factor Sox. Several Sox transcription factors have been linked to neural crest evolution and development [37]. Finally, rs12067906 (p-value = 9.584×10^{-6}) also maps in a DNAse hypersensitive region and onto two TFBS (corresponding to Gfi1 and HLF, both related to leukemia and several related pathologies). In addition, this site is bound by the protein RAD21, involved in DNA repair.

**Discussion**

Understanding the molecular and cellular processes required for proper ENS development, and therefore the corresponding defects that lead to HSCR, requires of the knowledge of the functionalities affected by the genes affected in the disease.

The PBA strategy has already been successfully applied to the study of some traits such as coronary heart disease risk [21], bipolar disorder [22], Crohn's disease, hypertension, rheumatoid arthritis or diabetes among others [23]. These studies led to the identification of numerous pathways implicated in disease predisposition that would have not been revealed using standard single-locus GWAS statistical analysis criteria. Many of such pathways had long been assumed to contain polymorphic genes that lead to disease predisposition. The same conclusions can be extracted from our results that reveal a clear association of GO terms connected to Ras signalling, widely known as a pathway with a key role in ENS formation.

Regarding network analysis, it exploits the information contained in the interactome with the idea that proteins close in the interaction network will have a higher probability of causing the same disease and constitutes a powerful technique to detect gene-disease associations. Network analysis has been successfully applied to discover genes in different diseases, such as ataxias [38], Huntington disease [39], schizophrenia [40] or Alzheimer's [41]. Therefore we considered it a really useful tool to be applied to the results derived from the multinational GWAS in the context of HSCR.

The recent description of HSCR-associated functional modules in Spanish population [17] constitutes an excellent initial functional hypothesis of the molecular mechanism of HSCR. We have used all the data available of an international cohort of 162 trios of short-segment Hirschsprung's disease to carry out a multi-population PBA study that validate the initial functional hypothesis. Specifically, our study has led us to the confirmation of a spectrum of different GOs related to the disease, being of special interest those terms related to signal transduction, such as the ones connected to Ras signaling. The Ras pathway is known to be one of the intracellular signaling mediated by the RET receptor, and is involved in cell survival and proliferation, both of them key biological processes related to ENS formation [42]. Thus, in spite of some apparent bias due to the size of the genes in these GO terms, there are solid biological basis for the involvement of such pathways in the disease. In addition, previous studies have demonstrated that signaling through the small Rho GTPases is also important for colonization of the gut by enteric neural crest cells and the concomitant growth of axons [42]. These results strongly suggest that members of the Ras/Rho protein signal transduction or regulators may play a key role in the

| SNP            | Nominal p-value | Adj. p-value (BH) | Chip | RefSeq genes | Feature                                | Proteins bound       | Motifs changed |
|----------------|-----------------|-------------------|------|--------------|----------------------------------------|----------------------|----------------|
| rs2435367      | 8.067×10^{-11}  |                   | Sty  | 5.9 kb 5' of RET | DNAse hypersensitivity                 |                      |                |
| rs2505526      | 8.353×10^{-7}   | 0.01506847        | Sty  | 7.5 kb 3' of RASGEF1A | Ultraconserved region [35]         | Foxj2                |                |
| rs16838932     | 4.459×10^{-7}   | 0.021091873       | Nsp  | 421 kb 5' of SLC39A10 | Ultraconserved region [35]         |                      |                |
| rs2659635      | 7.098×10^{-6}   | 0.08204207        | Sty  | 125 kb 5' of XRCC6BP1    | Ultraconserved region [35,36]       | Sox                  |                |
| rs4570660      | 7.744×10^{-6}   | 0.08479448        | Sty  | 65 kb 3' of APOBEC1    | DNAse hypersensitivity              |                      |                |
| rs12067906     | 9.584×10^{-6}   | 0.0994469         | Sty  | 96 kb 5' of RGS21     | DNAse hypersensitivity              | RAD21, Gfi1, HLF    |                |

In this case all the trios have been analysed together for each chip.
pathogenesis of HSCR (Figure 2). Previous studies have proposed some specific genes included in these GOs as potential candidate genes for HSCR. For instance, it has been shown an under-expression of the gene Arhgef3 in mice deficient for RET when compared to wildtype mice, which suggests its role in ENS formation. Interestingly, the human homologue for this gene, ARHGEF3, maps to 3p14, a chromosomal region previously described as a susceptibility locus for HSCR [11,43], although to date its candidacy has not been further evaluated. On the other hand, migration of enteric neural crest cells in the gut wall during embryogenesis requires interactions between the migrating neural crest cells and the extracellular matrix (ECM) environment in different regions of the developing gut. It implies a key role of cell migration during enteric nervous system formation, and would support the association to HSCR obtained for several related GO terms. Therefore, all these associated GOs might provide potential candidate genes implicated in ENS formation and also the pathogenesis of HSCR.

Our findings clearly show that, while the genes most associated to HSCR are essentially different in the five analyzed populations, gene modules with common functions (GO terms) are the same. Thus, the comparative analysis of the populations is revealing two important facts: a) the GO biological processes significantly associated to the disease in the different series of the ICHSCR strongly suggest that HSCR is caused in the different populations by different particular genes belonging to the same (or related) GO modules and b) such gene modules carry out biological functions that can be assimilated (using the GO hierarchy) to neurogenesis and signaling. The network analysis also points to the same processes. In other words, although we cannot exclude the existence of some causative genes common to all the populations, which still remain undetected, our results rather point to a scenario in which HSCR is the result of different genes causing approximately the same phenotypic effects in different populations.

Additionally, the observations made with the extra-genic SNPs suggest other possible disease mechanisms for HSCR more related to regulation or DNA instability. Some of these SNPs were markers of DNA hypersensitivity regions in the neighborhood of genes such as RET and, moreover, some of them directly point to TFBSs. One of the SNPs was pointing out a region of binding of a protein involved in DNA repair.

Conclusions

Independent evidences obtained from common gene functionality and from physical protein-protein interactions point to HSCR as a disease caused by variants in genes belonging to some GO modules related to neurogenesis, in particular within the context of ENS formation, and signaling. Moreover, the analysis of extra-genic SNPs in a functional context provided by the recent publication of the results of the ENCODE project [32] provides additional evidences in this direction. Interestingly, while the gene associations were different across populations the affected functionalities were always the same. This suggests that the known difficulty in validating genes in different populations [1,44,45] could be more a consequence of the multigenic nature of the disease that a sampling problem. In this scenario, the low percentage of the variance of traits explained by individual genes [46] is an obvious consequence of the fact that many complex diseases are the result of different combinations of variants that occur in different populations just by founder effects. Such different sets of variants collectively cause a malfunction of particular functional modules, which constitute the ultimate cause of the disease.

In summary, this comprehensive profile of functional modules (GO) has proven to be a useful resource for future developmental, biochemical and genetic studies. Our findings indicate that this approach can help to identify candidate genes for human disease susceptibility loci. Beyond technical considerations on the advantages of using functional modules in the analysis of genotype data, the biological pathways highlighted by our study provide insights into the complex nature of HSCR, opens new opportunities for validation of new disease genes and may help in the definition of relatively tractable targets for therapeutic intervention.

One known limitation of function-based approaches is that variants not mapping within or close to genomic elements with a functional annotation remain unused in the study. However, the recent availability of new functional domains provided by the ENCODE consortium will allow the extension of theses function-based studies (PBA, network analysis, etc.) beyond the conventional studies based on genes or known regulatory elements such as miRNAs.

Additional files

Additional file 1: Table S1. Summary of the effects of the different steps of the quality control.

Additional file 2: Figure S1. Summary of the effects of the different steps of the quality control.

Additional file 3: Figure S2. Boxplots of gene length distribution of genes lengths within GO terms. Significant GO terms from Table 2 are plotted in blue. The background distribution of gene lengths in the rest of non-significant GO terms is represented in yellow.

Additional file 4: Figure S3. Boxplots of gene length distribution of genes lengths within GO terms. The first boxplot on the left, in yellow, represents the distribution of genes in all the non-significant GO terms. The rest of boxplots in blue correspond to the significant GO terms from Table 2.
population [17] have also been added. Significant terms have been color-coded according the number of populations in which the GO terms was found to be significant. The darkest values corresponds to GO terms significant in five populations and the palest in only one population (see Additional file 6: Table S2).

Additional file 5: Figure S4. Tree hierarchy depicting the relationships between GO terms significantly associated to HSCR (FDR adjusted p-value < 0.05) using the PBA [25] as implemented in Babelomics [26] in the four country populations of the Consortium: French, Italian, Dutch and USA for the Sty chip. The results previously obtained for the Spanish population [17] have also been added. Significant terms have been color-coded according the number of populations in which the GO terms was found to be significant. The darkest values corresponds to GO terms significant in two populations and the palest in only one population (see Additional file 7: Table S3).

Additional file 6: Table S2. GO modules significantly associated to HSCR (FDR adjusted p-value < 0.05) using the PBA [25] as implemented in Babelomics [26] found in the analysis individualized by population in the Sty chip.

Additional file 7: Table S3. GO modules significantly associated to HSCR (FDR adjusted p-value < 0.05) using the PBA [25] as implemented in Babelomics [26] found in the analysis individualized by population in the Sty chip.

Additional file 8: Figure S5. Significant sub-network of 65 genes, allowing for one intermediate gene, associated to HSCR (p-value = 0.04) in the Sty chip. The network analysis [28] is carried out by using the functional analysis options (Set enrichment analysis/NetworkMiner) of Babelomics [26] on the list of genes ranked by the p-values obtained upon the application of a conventional TDT association test as implemented in PLINK [24].

### Acknowledgements

This work was supported by the Instituto de Salud Carlos III (ISCIII), Spanish Ministry of Economy and Competitiveness, Spain (PI10001290); Spanish Ministry of Economy and Competitiveness (BIO2011-27069), GVA-FEDER (PROMETEO/2010/001) and Consejeria de Innovación Ciencia y Empresa de Ministry of Economy and Competitiveness, Spain (PI1001290); Spanish Complex Disease Genomics, McKusick-Nathans Institute of Genetic Medicine, Johns Hopkins University School of Medicine, Baltimore, MD, USA.

### Author details

1. Department of Genetics, Reproduction and Fetal Medicine, Institute of Biomedicine of Seville (IBIS), University Hospital Virgen del Rocío/CSC/ University of Seville, Av. Manuel Siurot s/n, 41013 Seville, Spain. 2. Centre for Biomedical Network Research on Rare Diseases (CIBERER), Valencia, Spain. 3. Department of Computational Genomics, Centro de Investigación Príncipe Felipe (CIPF), C/Eduardo Primo Yllera, 3, 46012 Valencia, Spain. 4. Center for Complex Disease Genomics, McKusick-Nathans Institute of Genetic Medicine, Johns Hopkins University School of Medicine, Baltimore, MD, USA. 5. Department of Medical Genetics, University of Groningen, Groningen, The Netherlands. 6. INSERM U-781, AP-HP Hôpital Necker-Enfants Malades, Paris, France. 7. Laboratorio di Genetica Molecolare, Istituto Gaslini, Genova, Italy. 8. Functional Genomics Node (INB), CIPIF, Valencia, Spain.

### Received

23 May 2013 Accepted: 19 November 2013 Published: 2 December 2013

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