Integration of multigene panels for the diagnosis of hereditary retinal disorders using Next Generation Sequencing and bioinformatics approaches

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ABSTRACT

In recent years, Next-Generation Sequencing (NGS) opened a new way for the study of pathogenic mechanisms and for molecular diagnosis of inherited disorders. In the present work, we focused our attention on the inherited retinal dystrophies (IRDs), a group of specific disorders of the retina, displaying a very high clinical and genetic heterogeneity, whose genetic diagnosis is not easily feasible. It represents a paradigmatic example for the integration of clinical and molecular examination toward precision medicine.

In this paper, we discuss the use of targeted NGS resequencing of selected gene panels in a cohort of patients affected by IRDs. We tested the hypothesis to apply a selective approach based on a careful clinical examination. By this approach we reached a 66% overall detection rate for pathogenic variants, with a 52% diagnostic yield. Reduction of the efforts for

validation and classification of variants is a clear advantage for the management of genetic testing in a clinical setting.

INTRODUCTION

Inherited retinal dystrophies (IRDs) are a group of rare diseases due to a progressive degeneration of retinal photoreceptors, that can lead to vision loss [1,2]. IRDs comprise several different disorders characterized by clinical and genetic heterogeneity, often displaying a phenotypic overlap [3]. Many IRDs are characterized by progressive degeneration of both cone and rod photoreceptors, making the clinical differential diagnosis difficult, especially in the advanced stages [4].

Additionally, there are also syndromic forms in which retina is not the only affected tissue and additional organs can be involved, such as the Usher Syndrome (USH) and the Bardet-Biedl Syndrome (BBS). Furthermore, clinical symptoms can be progressive with variable onset and intra-familial variability, due to an incomplete penetrance and variable expressivity, making the clinical picture more complex [1]. All these factors often complicate or delay a precise diagnosis [1,2,5].

By a genetic point of view, IRDs displays locus and allelic heterogeneity [6], with more than 200 causative genes, that make the genetic characterization very complex. The advent of nextgeneration sequencing (NGS) has opened new frontiers in genetic diagnostics of IRDs, exploiting the high-throughput parallel sequencing and the simultaneous analysis of several samples. Indeed, the overall mutation detection rate for IRDs is variable [6], ranging from 36% to 60%, leaving many cases still genetically unsolved. More than 4000 pathogenic variants have been identified in causative genes, that can converge to the same phenotype [6] or can show different symptoms [1], complicating the molecular diagnosis. Lastly, since some IRDs causative genes are associated to specific inheritance traits (AR, AD, X-linked), a targeted genetic analysis could be more effective, although sometimes establishing the inheritance mode in an affected family is difficult [1].

Considering all the above, it is often complex to determine *a priori* which genes are to be analyzed and a "non-hypothesis-driven" approach has been applied in large NGS studies [7-9].

In the diagnostic laboratory, such an approach increases the risk to identify variants of uncertain significance, complicating the interpretation and implying a big effort in classification.

In this paper, we describe the strategy adopted by our multidisciplinary team to optimize the integration of clinical data and NGS targeted resequencing for the diagnostics of the different forms of IRDs. Our approach for the molecular diagnosis of IRDs, including genes that fit with the phenotype, allowed us to obtain a 66% overall mutation detection rate, consistent with the best rates obtained with the "non-hypothesisdriven" approach.

MATERIALS AND METHODS

Clinical diagnosis and sample collection

This investigation conformed to principles outlined in the Declaration of Helsinki.

We collected 35 unrelated affected patients with different forms of IRDs.

All patients underwent an ophthalmic evaluation at the Department of Ophthalmology of San Raffaele Hospital (Milan, Italy), including best corrected visual acuity by means of Early Treatment Diabetic Retinopathy (ETDRS) standard charts, biomicroscopy, color fundus photography, fundus autofluorescence, electrophysiological tests, and spectral-domain optical coherence tomography. Clinical and family history details were collected during genetic counseling interview. Written informed consent for genetic analysis was obtained from all subjects. Genetic analysis was performed at Laboratory of Clinical Molecular Biology of San Raffaele Hospital (Milan, Italy).

Genomic DNA (gDNA) was extracted from peripheral blood using the automated extractor Maxwell16 (Promega, Milano, Italy); the concentration and gDNA quality were determined using Qubit[®] Fluorometer (Thermo Fisher Scientific).

Library enrichment and sequencing

Sample enrichment and paired-end libraries preparation were performed using the commercial kit TruSight One (Illumina, San Diego, CA, USA), starting from 50ng gDNA, following the manufacturer's instructions (Document #100000006694 v00).

TruSight One Sequencing panel includes 4,813 genes associated with known clinical phenotypes, according to the Human Gene Mutation Database_HGMD (http://www.hgmd.cf.ac.uk/ ac/index.php), Online Mendelian Inheritance in Man, OMIM (www.omim.org), and GeneTests (www.genetests.org). The entire gene list is published on www.illumina.com (Pub. No. 0676-2013-016 current as of 04 January 2016). Sequencing was performed on NextSeq500 instrument (Illumina, San Diego, CA, USA) with a flow cell high output, 300 cycles PE (150 x 2).

NGS data analysis

The read alignment and variant calling were performed with BaseSpace Onsite Sequence Hub. For each case, the analysis of variants was focused on one or more gene panels based on the different clinical phenotypes. The variants were then annotated using Illumina VariantStudio data analysis software. For the identification of possible causative variants, filters were applied taking into account: 1) the quality parameter, 2) the MAF (Minor Allele Frequency) >2% in the 1000Genomes and ExAC database, 3) the localization of the variants, considering only the exonic and intronic regions at \pm 20 bp from the coding regions, to identify possible splice-site variations.

In order to optimize the data analysis process and to focus on the identification of causative variants, we created panels of disease genes associated to the different forms of IRDs, as reported in Table 1. Particularly, we set panels for nonsyndromic forms (Achromatopsia (ACHM); Best vitelliform macular dystrophy; Congenital stationary night blindness (CSNB); Choroideremia; Stargardt disease; Retinitis pigmentosa) and for syndromic forms (Bardet-Biedl S., Refsum disorder, Cohen S., Stickler S., Usher S.). We chose causative genes for each disease panel based on public databases, such as OMIM (http://www. ncbi.nlm.nih.gov/omim) or RetNet[™] (https:// sph.uth.edu/retnet/) and from the literature [4,6,10-12]. After primary analysis, the search for causative variants started by considering the panel of genes associated to the clinical suspicion. If the suspicion was less focused, more than one panel is analyzed.

Interpretation of putative variants was performed using Alamut[®] Visual (Interactive biosoftware), that integrate data from several databases, such as NCBI, UCSC, ClinVar, HGMD Professional, and *in silico* tools prediction, such as Polyphen, Sift, Mutation Taster. Candidate variants were classified according to the ACMG criteria in 5 categories:

- class 1: benign,
- class 2: likely benign,
- class 3: uncertain significance (VUS),
- class 4: likely pathogenic,
- class 5: pathogenic [13,14].

Analysis flow chart is reported in Figure 1.

Table 1

Different panels of disease genes associated to the different forms of IRDs

	Inherited Retinal Dystrophies											
		Ν	lon-Synd	romic Form	ıs		Syndromic forms					
	Achromatopsia	Best macular dystrophy	Congenital Stationary Night Blindness	Choroideremia	Stargardt disease- cone-rod dystrophy	Retinitis pigmentosa- rod-cone dystrophy	Bardet-Biedl Syndrome	Cohen Syndrome	Stickler Syndrome	Usher Syndrome	Refsum disease	
Orpha	49382	1243	215	180	827	791	110	193	828	886	773	
ICD-10	H53.5	H35.5	H53.6	H31.2	H35.5	H35.5	Q87.8	Q87.8	Q87.0	H35.5	G60.1	
Incidence	1-9 / 100 000	1-9 / 100 000	Unknown	1-9 / 100 000	1-5 / 10 000	1-5 / 10 000	1-9 / 1 000 000	Unknown	1-9 / 100 000	1-9 / 100 000	1-9 / 1 000 000	
Onset	Infancy, Neonatal	Childhood, Adolescent	Neonatal	Childhood, Adolescence, Adulthood	Childhood, Adolescence, Adulthood	Childhood, Adolescent, Adult	Prenatal, Neonatal, Childhood	Neonatal, Childhood	Childhood	Neonatal, Childhood	Infancy, Childhood, Adolescence, Adulthood	
Inheritance mode	AR	AD	AD; AR; X-linked	X-linked	AD; AR	AD; AR; X-linked; Mitochondrial inheritance	AR	AR	AR; AD	AR	AR	
Prevalence of mutations	75-90%	96%* (familial forms) 70%* (non familial forms)	95%*	95%*	65-70%	75%	90%*	70%*	100%*	80-85%*	100%*	
N. of genes of panel	7	3	14	1	43	63	18	1	5	11	2	
Genes	ATF6	BEST1	CABP4	СНМ	ABCA4	ABCA4	ARL6	VPS13B	COL11A1	ADGRV1	PEX7	
	CNGA3	IMPG2	CACNA1F		ADAM9	BBS1	BBS1		COL11A2	CDH23	РНҮН	
	CNGB3	PRPH2	CACNA2D4		AIPL1	BBS2	BBS10		COL2A1	CIB2		
	GNAT2		GNAT1		C2orf71	C2orf71	BBS12		COL9A1	CLRN1		
	PDE6C		GNB3		C8orf37	C8orf37	BBS2		COL9A2	HARS		
	PDE6H		GPR179		CABP4	BEST1	BBS4			MYO7A		
	RPGR		GRK1		CACNA1F	CA4	BBS5			PCDH15		
			GRM6		CACNA2D4	CDHR1	BBS7			PDZD7		
			NYX		CDH3	CERKL	BBS9			USH1C		
			PDE6B		CDHR1	CLRN1	CEP290			USH1G		
			RHO		CEP290	CNGA1	LZTFL1			USH2A		
			SAG		CERKL	CNGB1	MKKS					
			SLC24A1		CLN3	CRB1	MKS1					
			TRPM1		CNGA3	CRX	NPHP1					

		C1QTNF	CYP4V2	SDCCAG8		
		CNGB3	DHDDS	TRIM32		
		CNNM4	EYS	TTC8		
		CRB1	FAM161A	WDPCP		
		CRX	FLVCR1			
		CYP4V2	FSCN2			
		ELOVL4	GUCA1B			
		FSCN2	HGSNAT			
		GNAT2	IDH3B			
		GUCA1A	IMPDH1			
		GUCY2D	IMPG2			
		KCNV2	KLHL7			
		PDE6C	LRAT			
		PDE6H	MAK			
		PITPNM3	MERTK			
		PROM1	NR2E3			
		PRPH2	NRL			
		RAB28	PDE6A			
		RAX2	PDE6B			
		RDH12	PDE6G			
		RDH5	PRCD			
		RGS9	PROM1			
		RGS9BP	PRPF3			
		RIMS1	PRPF31			
		RP1L1	PRPF6			
		RPGR	PRPF8			
		RPGRIP1	PRPH2			
		SEMA4A	RBP3			
		TIMP3	RBP4			
			RDH12			
			RGR			
			RHO			
			RLBP1			
			ROM1			
			RP1			

			RP1L1			
			RP2			
			RP9			
			RPE65			
			RPGR			
			SAG			
			SEMA4A			
			SNRNP200			
			SPATA7			
			TOPORS			
			TTC8			
			TULP1			
			USH2A			
			ZNF513			

Data available on Orphanet (<u>http://www.orpha.net</u> - Last update: August 2017) and Genereviews (<u>https://www.ncbi.nlm.nih.gov/</u> books/NBK1116).

* Data reported on Genereview.

Identified variants were validated using Sanger Sequencing on AB3730 sequencer (Applied Biosystem), according to the manufacturers' protocols. (Primer and PCR conditions available on request). Moreover, in order to avoid undetected variants in regions with a low number of reads, all target regions of causative genes with a coverage <10X were analyzed by Sanger sequencing.

RESULTS AND DISCUSSION

Parameters of NGS raw data

All the 35 patients have been sequenced for 4813 genes, included in TruSight One panel (Illumina) using Illumina NextSeq500.

Runs had a mean cluster density equal to 217 k/ mm2. We obtained a mean read enrichment of 59% and target aligned read of 99%. The mean coverage for the analyzed genes associated to the different forms of IRDs was 300X.

Analysis and classification of detected variants

In our cohort, excluding common variants, we detected a total of 57 variants in 29 genes; 30 were novel and 27 were already reported in dbSNP as rare variants. In three patients no variants were found (9%), while the others (91%) presented with different variants with the exception of two pathogenic variants in *ABCA4* (NM_000350.2: c.5882G>A; NM_000350.2: c.5018+2T>C), identified in four different unrelated patients.

Considering all the detected variants, 66.7% (38/57) were missense, 10.5% (6/57) were stopgain, 7% (4/57) were frameshift changes, 8.8% (5/57) may alter splice sites, 1 variant was a startloss (1.8%), 1 was an in-frame insertion (1.8%), 1 was an in-frame deletion (1.8%) and 1 was a deletion of two whole exons (1.8%) (Figure 2). All the 57 variants were confirmed by Sanger sequencing or MLPA.

Figure 1Workflow of NGS analysis

1. LIBRARY ENRICH	MENT AND SEQUENCING			
Illumina TruSight One kit	Illumina NextSeq500 Instrument			
1				
2. NGS D	ATA ANALYSIS			
BaseSpace On Site	Variant Studio software			
1	Frequency <2%			
3. DAT	A FILTERING			
Selection of gene panel based on clinical diagnosis				
4. VARIANTS	SINTERPRETATION			
Alamut software	ACMG guidelines			
1				
5. IDENTIFICATION	OF CAUSATIVE VARIANTS			
YES -> REPORT	NO → Step 3 with a new gene panel based on a clinical reclassification			

The flow chart illustrates the main steps from the sequencing to the clinical report.

Among the 29 genes, the majority (22/29) present a single variant while seven genes are multivariated (Figure 3).

According with the ACMG guidelines [13], 11 variants were classified as pathogenic (class 5), 19 as likely pathogenic (class 4) and the remaining 27 as variants of unknown significance (VUS, class 3).

Evaluation of the diagnostic yield and genotype-phenotype correlation

We found pathogenic or likely pathogenic variants in 23/35 (66%) patients and consistent with the subject clinical presentation. Among these, we were able to reach the genetic diagnosis in 18/35 (52%) patients while in 5/35 (14%) patients we obtained only a partial diagnosis because of the detection of only one causative recessive variant. In 9/35 (26%) patients we identified heterozygous variants with unknown significance (VUS) in diseasegenes but in 5 of them the genotype did not fit to the disease inheritance manner and the genetic diagnosis remained incomplete. Finally, 3 patients were wild-type in analyzed causative genes. In these cases, a multidisciplinary rediscussion would be suggested in an attempt to define further testing or the potential for a research approach.

The majority of patients not reaching the genetic diagnosis had non-syndromic phenotypes, in particular two of the patients with no variants had a

Table 2 The p obtain	The percentance of complete, partial and total diagnostic yield obtained using our multigene panel approach for each disease									
Total patients = 35	Clinical	Patients	Complete	Partial	Total					
Overall diagnostic yield (%) = 51	phenotype	(n)	yield % (n)	yield % (n)	yield %					
	Pattern dystrophy	1	100 (1)		100					
	Bardet-Biedl S.	1	100 (1)		100					
	Best Disease	5	60 (3)		60					
Disease	Complex phenotype; retinal dystrophy (rod-cone or cone-rod)	11	36 (4)	18 (2)	54					
	Retinitis Pigmentosa	4	75 (3)		75					
	Stargardt disease	11	36 (4)	27 (3)	63					
	Stickler S.	1	100 (1)		100					
	Usher S.	1	100 (1)		100					

clinical diagnosis of Best, while in the case of retinal dystrophy, retinitis pigmentosa and Stargardt, a high proportion of patients had a partial or inconclusive diagnosis due to the presence of only one pathogenic variant or to the presence of VUS. In particular, for 5 patients with a partial diagnostic yield we can suspect the presence of a second pathogenic variant in a deep intronic region, as is the case for *ABCA4* or the presence of a structural variant not identified by sequencing. In Table 2, we reported the obtained diagnostic yield for each disease.

In Table 3 are listed all the genes with variants identified in the present work in association with different diseases. It is possible to appreciate

that the larger genetic overlap is between the retinal dystrophies and RP phenotypes (Table 3, the shaded lines).

In the present work, we applied a targeted NGS resequencing for genetic testing of IRDs; selection of gene panels was done based on the clinical suspicion (Table 1) allowing us to reduce the number of genes tested. We reached a diagnosis in a proportion of patients that was consistent with the results from other studies, where wider panels were used. Based on these findings, this approach, reducing the efforts needed for classification and validation of variants, seems to be more suited in the diagnostic field.



We identified 57 variants in 29 genes in our cohort and in the pie chart the percentage of each type of detected variant is reported.



Graph represents the number of detected variants (x-axis) for each gene (y-axis).

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Table 3	The gene	tic overlap	oping betwee	en retinal dy	strophies a	and RP phe	notypes
Pattern dystrophy	Bardet- Biedl Syndrome	Best Disease	Complex Phenotype; retinal dystrophy (rod-cone or cone-rod)	Retinitis Pigmentosa	Stargardt disease	Stickler Syndrome	Usher Syndrome
PRPH2	BBS4	BEST1	ABCA4	ABCA4	ABCA4	COL2A1	USH2A
		IMPG2	CDH23	C2ORF71	ATF6		
		PRPH2	CDHR1	CDHR1	CEP290		
			CEP290	CRB1	CRX		
			CNGA3	USH2A	GNAT2		
			CRB1		GPR98		
			FSCN2		PCDH15		
			IMPDH1		TOPORS		
			KCNV2				
			PDE6A				
			PDE6B				
			PITPNM3				
			PRPH2				
			RIMS1				
			RP1				
			RP1L1				
			RPGRIP1				

The coloured cells indicate the genes mutated in different clinical phenotypes.

CONCLUSION AND GENERAL REMARKS

Thanks to NGS, genetic testing costs are reducing rapidly with the potential for a broader access in the frame of health care systems. As NGS allows parallel analysis, it currently realizes a real improvement for personalized medicine, shortening the time needed to reach a diagnosis, nevertheless we still have to face a number of criticisms [15]. This report, showing an overall mutation detection rate for IRDs of approximately 60%, addresses the challenges ahead,

which include: a better understanding of the clinical significance of variants in disease genes; improvement of variant calling, especially for deep-intronic regions, regulatory sequences, promoters and structural variants (i.e.: extension of captured regions and improvement of tools for CNV detection); improvement of genotype-phenotype correlations and comprehension of more complex or not yet understood genetic mechanisms of diseases.

Correspondingly, the simultaneous sequencing of a large number of genes has resulted in increased detection of variants of unknown significance, which require interpretation for clinical purposes. The development of databases such as ClinVar and WES (Whole Exome Sequencing) variant allele frequency by ExAC Browser are gradually improving variant interpretation.

Similarly, programs such as SIFT, PolyPhen-2, and NNSPLICE are now widely used to predict the influence of a variant on protein localization, structure, and/or function. However, *in silico* predictions are not always consistent with functional studies and, despite recent advances, pathogenicity assessment remains challenging, particularly for hypomorphic, synonymous and non-coding variants. Ultimately, better tools are required, as well as improved knowledge of the genome and genome function.

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