METHODS

Human Mutation

Rapid and Efficient Human Mutation Detection using a Bench-Top Next-Generation DNA Sequencer



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ABSTRACT: Next-generation sequencing (NGS) technologies can be a boon to human mutation detection given their high throughput: consequently, many genes and samples may be simultaneously studied with high coverage for accurate detection of heterozygotes. In circumstances requiring the intensive study of a few genes, particularly in clinical applications, a rapid turn around is another desirable goal. To this end, we assessed the performance of the bench-top 454 GS Junior platform as an optimized solution for mutation detection by amplicon sequencing of three type 3 semaphorin genes SEMA3A, SEMA3C, and SEMA3D implicated in Hirschsprung disease (HSCR). We performed mutation detection on 39 PCR amplicons totaling 14,014 bp in 47 samples studied in pools of 12 samples. Each 10-hr run was able to generate ~75,000 reads and ~ 28 million high-quality bases at an average read length of 371 bp. The overall sequencing error was 0.26 changes per kb at a coverage depth of \geq 20 reads. Altogether, 37 sequence variants were found in this study of which 10 were unique to HSCR patients. We identified five missense mutations in these three genes that may potentially be involved in the pathogenesis of HSCR and need to be studied in larger patient samples.

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KEY WORDS: mutation detection; bench-top sequencer; HSCR; semaphorin

Introduction

The development of next-generation sequencing (NGS) technologies has been a boon to all genomics research; in particular, it has become important to human genetics, genome biology, and the understanding of human disease biology. Whole-genome [Wheeler et al., 2008], exome [Metzker, 2010; Schuster, 2008], and transcriptome [Durbin et al., 2010; Mardis, 2008] sequencing are becoming routine. The immense capacity (in excess of 30-50 Gb per run) and lengthy run times (longer than 1 week) of current sequencing systems have been used so far to assay the entire genome in a considerable number of samples to create reference datasets or for the inference of biological features on a genome-wide scale [Mardis, 2008; Metzker, 2010; Schuster, 2008]. One recent example is the 1000Genomes Pilot Project in which the genome sequences of 179 human samples were obtained at low coverage ($2-4\times$) to understand the patterns of rare and common human sequence variation in an unbiased manner [Durbin et al., 2010].

NGS technologies are quite error prone at the level of an individual sequence read so that accuracy is achieved by multiple read coverage of a variant base in an individual sample [Mardis, 2008; Metzker, 2010; Schuster, 2008] or across population samples with multiple occurrences of the same variant [Durbin et al., 2010]. Most genome sequencing projects have raw accuracies less than 99% [Drmanac et al., 2010; Ju et al., 2010; Mardis, 2008; Metzker, 2010], but exome sequencing for disease gene discovery has achieved higher accuracies (99.7%) [Ng et al., 2009, 2010] through higher coverage of the coding sequences of the genome. For some applications, such as comprehensive rare variant detection or identification of disease mutations, an even higher accuracy may be warranted. This higher accuracy is particularly demanded by clinical applications that generally target only a small set of genes relevant to the patient. Unfortunately, although greater accuracy can be achieved by increasing coverage depth, all current NGS platforms have capacities that are excessive for routine clinical applications. This suggests a need for smaller capacity next-generation sequencers that can accurately and rapidly sequence DNA for clinical applications.

The first NGS platform introduced was the Genome Sequencer FLX System from 454 Life Sciences (Roche) that used a highly parallel pyrosequencing system capable of producing ~400-600 million bases per 10-hr run [Margulies et al., 2005]. This technology was used to produce the first personal human genome sequence [Wheeler et al., 2008], that of James Watson, using sequence reads of 400-500 bp. It has been utilized in mutation detection studies as well [Bowne et al., 2010; Conrad et al., 2010; Kohlmann et al., 2010; Zaragoza et al., 2010]. However, the capacity of this sequencer exceeds the requirements of many small- and medium-scale targeted projects. Roche has recently introduced the GS Junior platform as a next-generation bench-top DNA sequencing solution scaled to suit the needs of small projects requiring a rapid turnaround time. With the analysis of 100,000 shotgun reads or 70,000 amplicon reads per run, together with a flexible sample pooling strategy using ligation multiplex identifiers (MIDs), the GS Junior might be one possible solution for rapid mutation detection and other similar applications. This machine has a maximum capacity of 35 million bases

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per run at an average read length of 400 bp and, because it employs the same chemistry, should be equivalent in performance to the previous GS FLX System.

Here we report on our experience in mutation detection for human disease gene discovery using the GS Junior system. Our laboratory has long been involved in the genetic analysis of Hirschsprung disease (HSCR; MIM# 142623), which is the most common genetic form of a functional intestinal obstruction in neonates [Chakravarti, 2001]. HSCR is a multifactorial neurocristopathy of the enteric nervous system and is associated with aganglionosis: the receptor tyrosine kinase RET plays a key role in all forms of HSCR and interacts with other genes to produce a variable phenotype [Amiel et al., 2008; Chakravarti, 2001]. Our recent studies have identified a locus on 7q21.11 containing significant association with HSCR with allelic effects independent of RET [S. Arnold et al., unpublished data, abstract 1311, ASHG annual meeting, November 4, 2010]. This locus contains three members of the type 3 semaphorin family of neuro-ligands that are attractive candidates for involvement in HSCR: SEMA3A (MIM# 603961), SEMA3C (MIM# 602645), and SEMA3D (MIM# 609907). Since the proteins encoded by these genes are closely related, we used the GS Junior NGS system to perform mutation detection to ascertain whether any or all of these three genes could contribute to HSCR. We report the successful parallel sequencing of pools of amplicons for comprehensive and accurate sequence analysis. Significantly, we show that potential mutations in all three genes may contribute to HSCR.

Materials and Methods

Samples Used

High-quality genomic DNA from 44 patients with HSCR was used for mutation detection in this study. Controls consisted of human genomic DNA (G1521: female; G1471: male) purchased from Promega Corporation (each corresponding to a mixture of six unrelated samples) and one HapMap reference sample (CEU, NA12814). Our patient samples do not have complete information on ancestry but the vast majorities are of European origin. All patient samples were obtained with written informed consent approved by the Johns Hopkins University School of Medicine IRB.

Amplicon Preparation

For DNA sequencing, we designed 39 amplicons (range: 233-606 bp; median: 360 bp) that were amplified using one of two methods: (1) Thermo-Start PCR Master Mix (AB-0938/15/DC/B): 1 µM of each primer (forward and reverse, primer sequences are available upon request), 25 µl of 2×Thermo-Start PCR Master Mix, 50 ng of DNA, and sterilized distilled water up to 50 µl for PCR amplification at the following conditions: 95°C for 15 min, 35 cycles of 95°C for 20 sec, 60°C for 30 sec, and 72°C for 1 min followed by 72°C for 5 min; (2) TaKaRa LA Taq (RR002M): 1 µM of each primer (forward and reverse, primer sequences are available upon request), 5 μl of 10×LA PCR Buffer II (Mg²⁺ plus), 8 μl of dNTP mixture (2.5 mM each), 0.5 µl TaKaRa LA Taq (5 units/µl), 50 ng of DNA, and sterilized distilled water up to 50 µl for PCR amplification at the following conditions: 94°C for 1 min, 30 cycles of 94°C for 30 sec, 55°C for 30 sec, and 68°C for 30 sec followed by 72°C for 10 min. PCR products were visualized on a 2.0% agarose gel by electrophoresis and purified with QIAquick PCR purification kit (QIAGEN, Valencia, CA). Subsequently, all amplicons derived from an individual's DNA sample were pooled in a length-weighted equi-

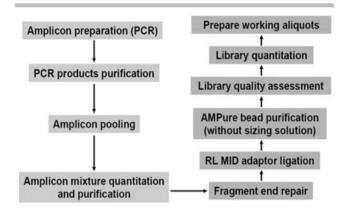


Figure 1. Workflow for preparing the amplicon library.

volume ratio (3 μ l for 200- to 250-bp products, 3.5 μ l for 251- to 300-bp products, 4 μ l for 301- to 350-bp products, 4.5 μ l for 351- to 400-bp products, 5 μ l for 401- to 500-bp products, and 6 μ l for 550- to 600-bp products). The pooled sample concentrations were measured by Nanodrop. Finally, 500 ng of each pool was purified with MinElute PCR purification kit (QIAGEN) and eluted in 16 μ l TE buffer.

Amplicon Sequencing

The sequencing library preparation was performed following the Rapid Library Preparation Method Manual (revised June 2010) with the following modifications: (1) the protocol was started at the fragment end repair step; (2) RL1-12 MID adaptors were ligated (we used three pools of 12 samples each and one pool of 11 samples); (3) during the AMPure XP purification step no sizing solution was used. Based on the individual sample concentration, the DNA libraries were diluted to 1×10^7 molecules/µl stock solution (Fig. 1). For the emulsion PCR (emPCR), up to 12 libraries were pooled in equimolar amounts and processed following the emPCR Amplification Method Manual (Lib-L, August 2010). The protocol was modified to take account of the amplicon length variation by (1) reducing the amount of amplification primer by half, and (2) using a low copy per bead ratio (0.3). The GS Junior Titanium Sequencing Kit and the Sequencing Method Manual (rev. June 2010) were used for DNA sequencing on a GS Junior Titanium PicoTiterPlate (PTP).

Sanger Sequencing

For verification, five purified amplicons from seven samples, each containing a newly detected variant, were sequenced by using the $1 \times$ BigDye Ready Reaction Mix (Applied Biosystems) on an Applied Biosystems 3730xl DNA Analyzer. Sanger data were analyzed using Sequencher version 4.10.1.

Allele-Specific PCR (ASP) Analysis

To validate a 4-bp deletion, allele-specific PCR (ASP) was performed. Selective amplification was achieved by designing two primer pairs, one each that matched the reference and variant allele. Genotyping was performed using the Thermo-Start PCR Master Mix and the same conditions as described above for the $10-\mu$ l PCR reaction. The results were visualized after running the samples in a 2% agarose gel (Supp. Fig. S1). The primers used were: reference/ wild-type primer (forward 5''GGAAGACCGATATCAAAGGTTC3' and reverse 5'GTTCAGTGTGCAGCTGTCCT3'); variant/assay primer (forward 5'GGAAGACCGATATCAAAGGTTG3' and reverse 5'GTTCAGTGTGCAGCTGTCCT3').

Mapping, Variant Identification, and Sequencing Accuracy

Two approaches were utilized for computational analysis of all GS Junior runs: 454's GS Amplicon Variant Analyzer version 2.5 (AVA) and tools available in Galaxy (http://main.g2.bx.psu.edu/) [Blankenberg et al., 2010; Goecks et al., 2010].

Analysis using AVA

We used the graphical user interface (GUI) for easy visualization of data. To enable the use of Rapid Library IDs we opted to initiate the GUI using an extra argument when opening the program:./gsAmplicon—enable "sequenceBlueprint;extraProjInit." The input files required to run AVA are the sff, the amplicon sequence, and the primer sequence files. AVA examines each read for the presence of either one of the primer sequences to assign each one to an amplicon. Once it identifies the amplicon to which the read belongs, the read is aligned only to that amplicon. The primer sequences are subsequently trimmed and substitutions, insertions, and deletions identified. Only variants found in both forward and reverse traces and present in at least 35% of all reads covering their respective base positions were further considered in our study.

As AVA GUI does not automatically provide a result file for coverage depth, we utilized the AVA Command-Line Interface (CLI) program, in addition to an in house-developed shell and MATLAB (http://www.mathworks.com/products/matlab/) scripts to look at depth per amplicon. The programing codes and scripts are provided in the Supporting Information.

Analysis using Galaxy tools

We used the Galaxy browser as a second approach to the analysis of sequence data. We generated a workflow (URL provided as Supporting Information) based on the information provided in the "454 Mapping: Single End" tutorial. Prior to uploading data to Galaxy we used 454's sff tools to extract sff files for each MID in a run. The Galaxy workflow is as follows:

- (1) Upload sff file for each individual.
- (2) Extract FASTA sequences from the sff file using sff converter tool.
- (3) Map reads to the hg19 reference genome using LASTZ mapper version 1.01.88 with the Roche-454 98% identity mapping mode.
- (4) Count mapped reads.
- (5) Filter uniquely mapped reads.
- (6) Extract the mapping information for all the uniquely mapped reads.
- (7) Convert the output from SAM format to BAM format using SAMtools Version 0.1.12.
- (8) Create a simple pileup from the BAM file using SAMtools.
- (9) Filter the pileup using a set depth per base value.

Post-Galaxy analysis for calling genotypes and calculating sequencing accuracy was performed using custom MATLAB scripts that are provided in Supporting Information.

Calling genotypes

To call a genotype at a particular base position two thresholds had to be met: minimum coverage (depth) per base (*N*) and minimum percentage of the variant allele or genotype-calling threshold (*T*). For variant calling at each position, we computed $t = (k/n)^*100$ where *k* is the count of nonreference alleles and n is the total depth. We considered only those depths that exceed our set threshold (i.e., $n \ge N$) and called genotypes by the set threshold as: call genotype AA, AB, and BB whenever $t \le T$, T < t < 100 - T, and $t \ge 100 - T$, where A and B are the reference and variant allele, respectively.

Calculation of sequencing accuracy

Genotypes obtained from samples sequenced in duplicate were compared to each other to assess sequencing accuracy. For optimization of sequencing accuracy, varying values for N and T were utilized.

Results

We sequenced the three genes *SEMA3A*, *SEMA3C*, and *SEMA3D* representative of the HSCR candidate locus on human chromosome 7q21.11. Each gene has 17 coding exons and is similar in cDNA sequence to the others (identity \sim 58% by CLUSTAL 2.1 multiple sequence alignment). We divided the total sequencing target of 14,014 bp into 39 amplicons varying in length from 233 bp to 606 bp.

To assess coverage and accuracy, we analyzed 12 samples per run, that is, 47 samples were divided into four sequencing runs (1-4). Each sample was marked with an identity tag (MID) so that its sequence could be extracted from all reads within the run. After checking coverage for each amplicon across all the samples in the first three runs, there were 16 amplicons across 35 samples that had low coverage. We repeated sequencing for these 16 amplicons for all 35 samples in runs 5 through 7. For run 4, a strong optimization was performed to obtain more uniform coverage for long amplicons than was achieved in the first three runs. Specifically, the lengthweighted equi-volume ratios were increased to 5.5 µl for 351- to 400-bp products, 8 µl for 401- to 500-bp products, and 12 µl for 550- to 600-bp products. Table 1 summarizes the following for the seven sequencing runs: number of filtered reads per run, total length of sequence data produced, average read length, and average sequence coverage at each base across all runs.

A number of features are evident from these data. First, individual samples were covered approximately uniformly within a pool: Figure 2 shows run no. 4 as an example. Second, the generated data allowed sensitive detection of variants with a median of 6,507 high-quality sequencing reads per individual. The average length of reads ranged between 313 bp and 435 bp and a median of 2.29 mb was sequenced per sample. Third, replication data for 16 amplicons across 35 samples enabled us to examine the sequencing error across two runs by using different combinations of depth per base (N) and genotype-calling thresholds (T) (Table 2). Based on these calculations, we chose a minimum depth of 20 reads per base and 35/65% as an optimum genotyping threshold to obtain an average error of 0.26 (range: 0.13–0.40) per kb.

One significant limitation of pyrosequencing is its apparent inability to correctly determine the number of bases within a homopolymeric stretch [Brockman et al., 2008]. Consequently, we paid particular attention to the resolution of homopolymeric and dinucleotide stretches. The ~14-kb sequence across 39 amplicons had 36 homopolymeric stretches (repeat \geq 6) and one dinucleotide

Table L. Sequencing Statistics	Table 1.	Seque	ncing	Statistics
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Run no.	No. of samples	Total no. of filtered reads	Total sequence generated (Mb)	Average length (bp)	Average sequence depth/base
1	11	72,191	23.29	323	151×
2	12	75,424	23.66	314	$141 \times$
3	12	84,441	27.44	325	163×
4	12	101,395	34.39	339	$205 \times$
5	12	60,243	26.25	436	156×
6	12	70,406	29.91	425	$178 \times$
7	11	71,587	31.06	434	202×

For each run, we show the number of human samples in the amplicon pool, the total number of filtered reads, and the total sequence generated to obtain the average read length and the sequence coverage as indicated.

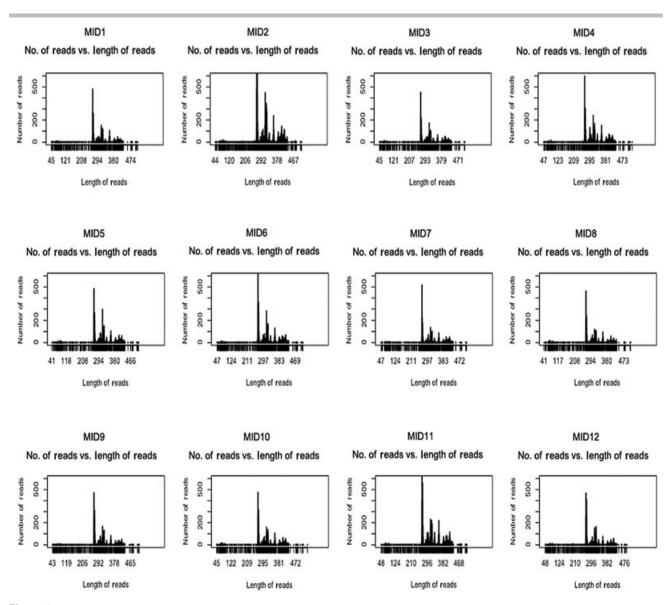


Figure 2. Distribution of read lengths in amplicon sequencing run no. 4. Each box represents the reads for one individual sample tagged with a specific multiplex identifier (MID).

stretch (repeat units \geq 6). Of these, 35 were resolved well; a specific example is shown in Supp. Figure S2. However, the two remaining features, a (AT)₁₉ dinucleotide and a T₁₃ homopolymer, which were close to a reverse and a forward sequencing primer, respectively, performed less optimally with a read depth of fewer than 50 (Supp. Fig. S3).

To assess the accuracy of insertion or deletion (indel) calls, we validated a 4-bp deletion polymorphism (AGAA, rs3832523) in intron 11 of *SEMA3A* through an ASP test on these samples. For detection of the variant B allele, the threshold to call the genotype was set as 35/65%. These values did not apply to the control samples (no. 18, 19) since each was a mixture of at least six samples. Of the

Table 2. Sequencing Error as Changes Per kb

	Heteroz	zygote discrimination t	hreshold
Sequencing depth	35%	30%	25%
All	1.02 (0.59–1.27)	1.27 (0.84–1.54)	1.37 (0.87–1.7)
>5×	0.49 (0.19-0.65)	0.60 (0.30-0.81)	0.74 (0.30-0.97)
>10×	0.41 (0.22-0.58)	0.49 (0.31-0.65)	0.61 (0.31-0.95)
>15×	0.37 (0.22-0.48)	0.45 (0.28-0.59)	0.54 (0.25-0.74)
>20×	0.26 (0.13-0.40)	0.34 (0.17–0.49)	0.40 (0.10-0.69)

Replication data for 35 samples and 16 amplicons enabled us to estimate the sequencing error across two runs as the fraction of discordant calls by using different thresholds of sequencing depth (N > 0, 5, 10, 15, and 20 reads) and heterozygote discrimination (T = 35%, 30%, 25%). Average error rates and their ranges are shown.

other 44 samples examined by both GS Junior and ASP, we obtained a concordance of 97.7% (43/44) (Table 3). Note that we identified no variant calls in genotypically identified reference homozygotes and obtained >90% (average: 96.9%) concordance for genotypically identified variant homozygotes. In contrast, the rate of sequencing concordance for 14 genotypically identified variant heterozygotes (excluding no. 18, 19) was between 18% and 54%. Of these, only one was an outlier at 18%, the remainder ranged from 39% to 54% (average: 47.2%). These results show the robustness of NGS for detecting variation, in heterozygotes in particular, given 100 or more reads. The sole failure occurred in a genotyped heterozygote that showed 18% variant calls among 252 reads. The 18% value is too large to be dismissed as a false positive and is likely from differential amplification of the normal and deleted alleles.

As a final comparison of accuracy, we included three samples that had been previously examined for variant detection in *SEMA3A* and *SEMA3D* by Sanger sequencing. As shown in Table 4, we obtained only one discordant call among 23 comparisons of eight coding variants at the standard 35/65% variant detection threshold by AVA. However, the variant reads were 0% of the forward (7 reads in total) and 79.31% of the reverse reads (29 reads in total). This

Table 3. Comparison of Genotypes of a 4-bp Deletion Variant Based on GS Junior Sequencing and Allele-Specific PCR (ASP)

Sample no.	Sample ID	Genotype (ASP)	Percentage of variant/total no. of reads	No. of F:R reads	Agreement
1	122.7	+-	51%/106	46:60	Yes
2	359.3	+-	53%/200	90:110	Yes
3	47.3	+-	39%/178	87:91	Yes
4	384.3		94%/100	53:47	Yes
5	392.3	+-	48%/125	67:58	Yes
6	346.3	+-	47%/221	104:117	Yes
7	242.4		99%/243	134:109	Yes
8	423.3	+-	47%/167	76:91	Yes
9	443.3	+-	46%/191	116:75	Yes
10	446.3	+-	54%/24	15:9	Yes
11	452.3		98%/288	132:156	Yes
12	434.3	+-	42%/163	89:74	Yes
13	432.3	+-	42%/191	94:97	Yes
14	416.3	+-	47%/151	79:72	Yes
15	411.3	+-	48%/441	211:230	Yes
16	399.3	+-	18%/252	113:139	No
17	402.2	+-	50%/157	85:72	Yes
18	Female	+-	16%/272	161:111	Pooled sample
19	Male	+-	15%/330	163:167	Pooled sample
20	63.3	++	0%/202	99:103	Yes
21	150.3	++	0%/133	67:66	Yes
22	252.3	++	0%/92	51:41	Yes
23	300.3	++	0%/372	175:197	Yes
24	348.3	++	0%/117	49:68	Yes
25	354.3	++	0%/330	158:172	Yes
26	355.3	++	0%/174	97:77	Yes
27	369.3	++	0%/344	174:170	Yes
28	370.3	++	0%/107	46:61	Yes
29	372.3	++	0%/126	71:55	Yes
30	406.3	++	0%/429	216:213	Yes
31	407.3	++	0%/180	86:94	Yes
32	408.3	++	0%/185	92:93	Yes
33	413.3	++	0%/133	75:58	Yes
34	422.3	++	0%/140	68:72	Yes
35	429.3	++	0%/293	175:118	Yes
36	435.3	++	0%/188	97:91	Yes
37	439.3	++	0%/151	82:69	Yes
38	440.3	++	0%/136	63:73	Yes
39	441.3	++	0%/124	60:64	Yes
40	444.3	++	0%/304	156:148	Yes
41	448.3	++	0%/419	222:197	Yes
42	449.3	++	0%/115	62:53	Yes
43	450.3	++	0%/266	118:148	Yes
44	451.3	++	0%/213	109:104	Yes
45	398.2	++	0%/200	105:95	Yes
46	NA12814	++	0%/525	263:262	Yes

The sequencing results are shown as % variant of total number of reads and the number of forward: reverse reads. Reference homozygote, heterozygote, and variant homozygote are represented as ++, +-, and - -, respectively. Altogether 44 comparisons were performed; one disagreement between the two methods is highlighted in bold.

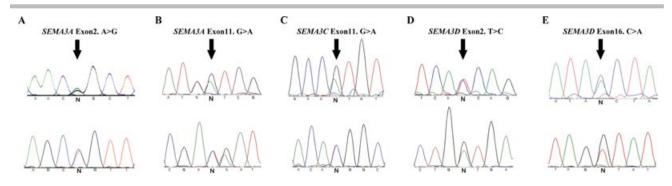


Figure 3. Sanger-sequencing validation of five rare missense heterozygote mutations. The five chromatograms are shown in (A) through (E) with forward and reverse direction sequencing results in the first and second rows, respectively; mutation locations are indicated by arrows.

Table 4.	Comparison	between	GS Junior	[•] and Sange	er Sequencing
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			Agreement betwee	en GS Junior and Sanger sequenc	ing
Gene	Location	Variant	Sample #335.3	Sample #398.2	Sample #402.2
	Exon2	121:A/G	Yes (AG)	Yes (AA)	Yes (AA)
	Exon2	228:A/G	Yes (AA)	Yes (AG)	Yes (AA)
SEMA3A	Exon11	250:T/C	Yes (CT)	Yes (TT)	Yes (CT)
	Exon11	251:G/A	Yes (GG)	Yes (GG)	Yes (AG)
	Exon17	365:A/G	No (Junior: AG; Sanger: GG)	Yes (AG)	Yes (AG)
	Exon2	265:T/C	Not examined	Yes (TT)	Yes (TT)
SEMA3D	Exon14	146:G/A	Yes (GG)	Yes (GG)	Yes (AG)
	Exon16	172:C/A	Yes (CC)	Yes (CC)	Yes (AC)

Three samples, which had been previously examined for variation in SEMA3A and SEMA3D by Sanger sequencing, were used as positive controls. Among eight coding-region variants across three samples there is one disagreement (highlighted in bold). Genotypes called are listed in parentheses.

suggests that we need greater experimental experience to set these variant detection thresholds to minimize false positives and false negatives. In other words, strand bias should also be considered in addition to the *N* and *T* thresholds for heterozygous detection. Finally, we successfully validated the five rare missense mutations in *SEMA3A*, *SEMA3C*, and *SEMA3D* by Sanger sequencing (Fig. 3).

In this study of 47 samples, we identified 37 variants of which 16 were coding and the remaining 21 in untranslated or intronic segments. The details of all detected variants, the majority of which have been observed in control but not diseaseor locus-specific databases, are provided in Table 5. Of relevance to HSCR are six of the 16 coding variants that were nonsynonymous: SEMA3A: c.160A>G (p.Ser54Gly); c.1303G>A (p.Val435Ile); SEMA3C: c.1009G>A (p.Val337Met); SEMA3D: c.193T>C (p.Ser65Pro); c.1843C>A (p.Pro615Thr), and c.2101A>C (p.Lys701Gln). Nucleotide numbering of the exonic variants reflects cDNA numbering with +1 corresponding to the A of the ATG translation initiation codon in the reference sequence, specifically, RefSeq NM_006080.2 for SEMA3A, RefSeq NM_006379.3 for SEMA3C, and RefSeq NM_152754.2 for SEMA3D. Of these, the SEMA3D K701O variation is a common polymorphism with a variant residue (Q) frequency of 0.28 in HSCR and an identical frequency in 1000Genomes samples. In addition, the SEMA3C V337M variation is observed in one HSCR patient and one of the control samples, with a 0.02 frequency in 1000Genomes samples. The SEMA3A S54G variant was not observed in either the HapMap exome sequencing project or the 1000Genomes project, while the remaining three changes, SEMA3A V435I, SEMA3D S65P, and SEMA3D P615T have all been observed as sequence alterations in the 1000Genomes samples and have allele frequencies of 0.009, 0.011, and 0.004, respectively. Interestingly, except for the SEMA3D

K701Q polymorphism, the other five alterations occur at highly conserved domains, either recognized (Sema and Ig domains) or not, and are predicted to be either damaging. Indeed, the residues in question are conserved across all mammals, other vertebrates, and the zebrafish, and the few exceptions (chicken residue T at SEMA3D S65P and zebrafish residue S at SEMA3D P615T) suggest sequence errors in the genome sequences from which these protein translations have been inferred (Supp. Fig. S4). In other words, we suspect that these five missense alterations have some role in HSCR, which requires follow-up in larger numbers of patients.

Discussion

Detection of DNA sequence variants is a central task in human genomic and genetic studies, and NGS technologies are capable of overcoming the many limitations inherent in Sanger sequencing [Galan et al., 2010; Lank et al., 2010; Lind et al., 2010; Schuster, 2008; Taudien et al., 2010]. As we show in this study, one individual can optimize and produce high-quality data on mutation detection in a short period of time using the bench-top GS Junior sequencer. At the same time, it is important to note that depending on the input, amplicon or shotgun libraries, the pre-GS Junior steps can be very labor intensive. The actual GS Junior protocols consist of (1) library preparation step (4 hr for 12 libraries), (2) emulsion PCR setup (1 hr) and emPCR amplification (5.5 hr), (3) breaking of emPCR and enrichment (2.5 hr), (4) sequencing setup (1.5 hr), and (5) sequencing run (10 hr). The completion of all steps requires 2 days for an individual experimenter. The system includes a computer preinstalled with GUI and CLI software, so that researchers can easily view their run information, assemble sequences, map reads to a reference genome, and analyze the amplicon data. Each program is

notation	
Variant an	
Table 5.	

	А	IJ	SEMA3C §	Synonymous	18/18	TAT	TAC	Υ	Υ	708	rs1949971	NA	NA	Yes	Yes	Yes	No	No
	Т	C	SEMA3C §	Synonymous	18/18	CCA	CCG	Р	Ь	676	rs1949972	NA	NA	Yes	Yes	Yes	No	No
	Т	C	SEMA3C	Intronic	ı		ı	ı			rs1405743	1		No	No	Yes	No	ī
	Т	U	SEMA3C	Intronic	1			ı	ı		rs2886793			No	No	Yes	No	
	IJ	C	SEMA3C 5	Synonymous	17/18	GTC	GTG	Λ	Λ	579	rs2272351	NA	NA	Yes	No	Yes	No	Ig
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The columns shown in order are: chromosome no. of the gene; hg19 nucleotide position; reference/variant allele on the + strand of genomic DNA; gene name; variant type (synonymous, missense, intronic); exon no;; reference codon; mutant codon; reference rodon; mutant codon; reference rodon; mutant residue position; dbSNP rslD; Pohyben2 prediction (using options: HumVar, hg19, All, Canonical); S117 prediction; Ng3 HapApa exomes Ng et al., 2009] (Yes = present); conservation to zebrafish (UCSC in algument residue position; dbSNP rslD; Pohyben2 prediction (using options: HumVar, hg19, All, Canonical); S117 prediction; Ng3 HapApa exomes Ng et al., 2009] (Yes = present); conservation to zebrafish (UCSC in algument arcoss Human, Chimp, Dog, Cow, Mouse, Rat, Chicken, and Zebrafish); 1000G variant (Yes = present in 8-4-2010 release); HGMD variant (Yes = present in professional version); protein domain from HPRD. Missense mutations are given in algument prion of the exonic variants reflects CDNA numbering with +1 corresponding to the AIG translation initiation codon in the reference sequence, specifically; RefSeq NM_006080.2 for *SEMA3A*, RefSeq NM_06539.3 for *SEMA3A*, and RefSeq NM_152754.2 for *SEMA3A*.

relatively simple to understand and returns output within minutes. The ability to use the GUI also enables experimenters to readily analyze various aspects of data, regardless of their computational prowess.

In this study, we evaluated whether parallel sequencing on the GS Junior system is suitable for mutation detection for disease gene discovery. We used a multiplex bar coded amplicon sequencing approach for three type 3 semaphorin family genes as an example. To enable uniform coverage of all amplicon targets, we introduced three modifications. First, since PCR favors amplification of smaller fragments in a complex mixture of different length templates, we modified the sequencing protocol by pooling amplicons relative to their size. Second, normal emulsion PCR amplification protocols with short amplicons (<400 bp) may result in an excessive number of amplified targets on the capture beads, thereby increasing signal intensity during incorporation as well as rapid consumption of the four nucleotide flow reagents during sequencing. To overcome this, we reduced the volume of amplification primer in the emulsion PCR from 80 µl to 40 µl. Third, it has been reported that when the DNAto-bead ratio is small and covers an optimal range, one obtains a linear relationship with the final enrichment percentage. Thus, we used a low copy per bead ratio (0.3) during the emulsion PCR amplification step since imprecise (±2-fold) library quantification can still give satisfactory results when the copy per bead ratio is low rather than high [Zheng et al., 2010].

Our experience suggests that all potential variants observed be given careful scrutiny with respect to base coverage, expected error rate, read length, bidirectional read support, and sequence context (homopolymeric and dinucleotide stretches). As observed in our experiments, read-coverage patterns varied across the targeted amplicons even after optimization. Other sources of variability include differential adapter-to-target fragments ligation, unequal PCR amplification efficiencies during library generation, and variations in amplicon size and GC content [Harismendy et al., 2009; Shendure et al., 2005], not to mention differential amplications of the two alleles in a diploid. These sources of variability need not be a limitation of NGS since samples can be somewhat "over-sequenced" to achieve a desired coverage level and, consequently, reduced error rate. In terms of variant identification, different criteria have been used in NGS studies depending on the platform, software, and specific study goals. However, for clinical applications, the major hurdle in the use of NGS technologies is how to set a reliable coverage/error threshold for optimizing false positives and false negatives. In the current study, we required a read coverage of 20-fold for variant identification after combining guidelines for the Genome Sequencer FLX systems (Genome Sequencer System Application Note 5 2007), and our and others' experience with the GS FLX platform [De Leeneer et al., 2011]. With respect to read percentage required for variant identification, we set our criterion to >35% to minimize false-positives, in turn anticipating false-negatives. It has been shown that the Genome Sequencer FLX system encounters difficulties when sequencing homopolymeric regions of more than 3 bp [Bordoni et al., 2008], and such stretches turned out to be major sources of sequencing errors. With the newly developed Titanium technology and software, containing various quality filters to remove poor-quality sequence, longer strings of up to 6 bp could be resolved very well; one example is shown in this study. The per-base error rates from 454 pyrosequencing are believed to be comparable to those from Sanger sequencing [Huse et al., 2007]. We show that even at a coverage depth of 20, the sequencing error is between 0.13 and 0.40 changes per kb with an average of 0.26. Depending on the purpose one may require much greater coverage.

Our results suggest that, in addition to substitutions, small deletion variants (four bases) can be reliably detected. The genotyping disagreement between GS Junior and ASP for sample 399.3 with respect to the deletion variant should not be regarded as a contrary result, because the 18% variant frequency is unlikely due to a sequencing error but rather to unequal amplification between the normal and deleted alleles during the sequencing protocol. With respect to the genotype disagreement in sample 335.3 between GS Junior and Sanger, the combined 63.89% variant frequency suggests a heterozygote. We speculate that the inconsistency is probably owing to sample contamination during either amplicon pooling or the library construction step.

HSCR is a multifactorial disorder that displays a highly variable phenotype with variation in recurrence risk by gender, familiality, segment length of aganglionosis, and associated phenotypes. The reasons for much of this variation are largely unknown, although gene discovery has clarified some genotype-phenotype correlations [Emison et al., 2010]. We undertook this sequencing study to assess the role of three type 3 semaphorin genes within a locus on 7q21.11 with significant association with HSCR [Arnold et al., abstract 1311, ASHG annual meeting, November 4, 2010]. Here, by analyzing the coding sequence of SEMA3A, SEMA3C, and SEMA3D in 44 HSCR patients, we detected five missense mutations that are potentially involved in the pathogenesis of HSCR, although many more samples need to be analyzed to demonstrate statistical significance. Semaphorins constitute a large family of signaling molecules originally identified as axon guidance cues [Kolodkin, 1998; Tran et al., 2007]. Data from previous studies have suggested a role for members of the semaphorin family in neural crest cell development [Anderson et al., 2007; Berndt and Halloran, 2006; Lwigale and Bronner-Fraser, 2009; Yu and Moens, 2005], defects in the proliferation, migration, and/or differentiation of which might be a cause of HSCR. The mutations we detected can, thus, be probes for altered function in cellular and animal models of HSCR.

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Conflicts of Interest: This study arose from our independent analysis of the GS Junior sequencing system as a beta test site.

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