ORIGINAL ARTICLE



Assessment of Illumina[®] Human mtDNA Genome assay: workflow evaluation with development of analysis and interpretation guidelines

Viktorija Sukser¹ · Filip Rokić² · Lucija Barbarić¹ · Marina Korolija¹

Received: 16 October 2020 / Accepted: 8 January 2021 / Published online: 29 January 2021 © The Author(s), under exclusive licence to Springer-Verlag GmbH, DE part of Springer Nature 2021

Abstract

Mitochondrial DNA (mtDNA) is a small but significant part of the human genome, whose applicability potential has gradually increased with the advent of massively parallel sequencing (MPS) technology. Knowledge of the particular workflow, equipment, and reagents used, along with extensive usage of negative controls to monitor all preparation steps constitute the prerequisites for confident reporting of results. In this study, we performed an assessment of Illumina® Human mtDNA Genome assay on MiSeq FGxTM instrument. Through analysis of several types of negative controls, as well as mtDNA positive controls, we established thresholds for data analysis and interpretation, consisting of several components: minimum read depth (220 reads), minimum quality score (41), percentage of minor allele sufficient for analysis (3.0%), percentage of minor allele sufficient for interpretation (6.0%), and percentage of major allele sufficient for homoplasmic variant call (97.0%). Based on these criteria, we defined internal guidelines for analysis and interpretation of mtDNA results obtained by MPS. Our study shows that the whole mtDNA assay on MiSeq FGxTM produces repeatable and reproducible results, independent of the analyst, which are also concordant with Sanger-type sequencing results for mtDNA control region, as well as with MPS results produced by NextSeq®. Overall, established thresholds and interpretation guidelines were successfully applied for the sequencing of complete mitochondrial genomes from high-quality samples. The underlying principles and proposed methodology on the definition of internal laboratory guidelines for analysis and interpretation of MPS results may be applicable to similar MPS workflows, e.g. targeting good-quality samples in forensic genetics and molecular diagnostics.

Keywords MiSeq · Mitochondrial DNA · Nextera XT · Evaluation · Analysis thresholds

Introduction

For such a relatively small portion of the human genome, mitochondrial DNA (mtDNA) exhibits extraordinary variability and unique features. The size of the human mitochondrial genome approximates 16,569 base pairs (bp; length may slightly vary due to insertions and deletions), which is on a scale of about 1:200,000 compared to the nuclear DNA. Despite its diminutiveness, mtDNA is essential for cellular

Marina Korolija mkorolija@mup.hr energy production and, thus, presents a vital part of our genome. It is enclosed within double-layered membranes of the cell's energy factories-mitochondria. Due to its wellprotected location, as well as circular nature, and the fact that there may be as many as several thousand copies of mtDNA per one cell (as opposed to nuclear DNA, present only in two copies per cell), this small genome is more resistant to environmental conditions and degradation than nuclear DNA. Therefore, it may well be the only source of genetic information recoverable in some cases, and even though it may not be used for individual identification (as all maternal relatives have the same mitochondrial genome sequence, with tolerable variations in indels and heteroplasmies), it is certainly preferable to no result at all. The aforementioned characteristics have established mtDNA as a valuable source in many fields of science, such as evolutionary biology, molecular anthropology, forensics, etc. [1].

¹ Biology and Fibres Department, Forensic Science Centre "Ivan Vučetić", Ilica 335, 10000 Zagreb, Croatia

² Laboratory for Advanced Genomics, Division of Molecular Medicine, "Ruder Bošković" Institute, Bijenička cesta 54, 10000 Zagreb, Croatia

Until fairly recently, the only part of mtDNA extensively investigated was the control region (CR), approximately 1100 bp in length, encompassing the origin of replication, other regulatory elements, and hypervariable regions (or segments; HVS-I, HVS-II, and HVS-III). Most of the mitochondrial sequence variation is concentrated in HVS, and mtDNA CR analysis by Sanger-type sequencing (STS) has become the gold standard employed in routine forensic casework, where sample material is scarce and challenging to process for various reasons (degradation, inhibitors, etc.). However, CR equals only around 7% of the complete mitochondrial genome, and in cases of more common mitochondrial haplotypes, this information alone cannot provide the resolution sufficient for forensic purposes [2]. Therefore, sequencing of the entire mtDNA clearly has great value, as inter-individual variation comes to the fore by revealing all 16,569 bp length of genetic information. Besides ethical and legal issues which stem from accessing the coding region sequence, analysis of whole mitochondrial genomes was simply not feasible previously with the Sanger sequencing method, as it was costly, laborious, time-consuming, and nearly impossible to apply on a large scale-few studies endeavoured to employ STS to produce whole mtDNA data (e.g. [3, 4]). In addition, population samples usually contain an abundance of genetic material of high quality, whereas forensic casework samples rarely come in such a pristine state, meaning STS of whole mtDNA would be even more difficult in the latter case.

Over the recent years, we have witnessed great technological leaps that brought about the next generation of sequencing platforms and chemistries, or rather as it is more commonly called, the massively parallel sequencing (MPS). It has advanced research in many areas of biology, including forensic science [5], where the focus of forensic genetics is gradually shifting from allele length-based identification to sequence variants, enabling even better power of discrimination. The field is being transformed into forensic genomics, since the sequencing of entire genomes (nuclear and/or mitochondrial) is not an unachievable feat in routine laboratory workflow anymore. The true challenge is to assemble all steps of the sequencing protocol into a single workflow, suited for a particular study, with sequencing data analysis being a singular challenge on its own [6]. Analysis and reporting for forensic purposes rely on compliance with internationally agreed and prescribed guidelines; wherefore, the method needs to be evaluated through internal validation performed by each laboratory [5, 7, 8]. Current mtDNA guidelines [9, 10] have been updated to some extent to accommodate MPS methods, and will certainly undergo further refinements as more and more MPS data are generated. Various studies have already shown repeatability, reproducibility, concordance to STS data, and overall reliability of MPS assays for analysis of whole mtDNA [11–17]. However, their approach to data analysis and interpretation differed, with bioinformatics solutions encompassing commercially available software, free online software, in-house developed and tailored pipelines, along with almost as diverse threshold settings.

In this work, we evaluated Illumina® Human mtDNA Genome assay on MiSeq FGxTM benchtop sequencer, in conjunction with BaseSpace® Sequence Hub applications for mtDNA analysis (namely, mtDNA Variant Processor and mtDNA Variant Analyzer). The assay is based on Nextera® XT library preparation, which consists of target enrichment by long-range PCR (mtDNA amplified in two overlapping amplicons), fragmentation, and tagging (performed by Nextera® XT transposome), dual index barcoding, and subsequent library purification and normalization. Libraries are pooled, denatured, and diluted prior to loading on the instrument, to undergo paired-end sequencing-by-synthesis reactions. From there, it is natural to proceed with data analysis in Illumina's bioinformatics online platform, thus streamlining the workflow and enabling faster data processing. We present here our approach to setting analysis and interpretation thresholds for the whole mtDNA analysis workflow, as well as evaluation of the entire workflow. Internal interpretation guidelines were developed herein, defined by multiple components of the thresholds (encompassing read depth, allele percentages, and quality), but the underlying principles of the approach hold potential for wider application in other similar MPS workflows. Our aim was to establish a reliable system suitable for sequencing complete mitochondrial genomes from high-quality samples of the type to be used for population study (i.e. buccal swab samples and blood), which is one of the prerequisites for using mitochondrial sequence information for forensic purposes.

Materials and methods

Sample collection and plan of experiments

For the purpose of this study, reference samples were collected from 11 volunteers. All participants gave detailed informed consent. From each person, two types of samples were collected: buccal swabs (collected on Whatman[™] Sterile Omniswab, GE Healthcare, UK) and blood (collected on Whatman[™] FTA[™] Classic Cards, GE Healthcare, UK). DNA was extracted from buccal swabs using the EZ1® DNA Investigator® kit on EZ1® Advanced XL instrument (Qiagen, Hilden, Germany), following the manufacturer's instructions [18]. As for dried blood on FTA[™] Cards, QIAamp® DNA Micro Kit (Qiagen) was used for DNA extraction, also according to the manufacturer's instructions [19]. All DNA extracts were subsequently quantified on Qubit[™] 3.0 Fluorometer using Qubit[™] dsDNA High Sensitivity kit (Thermo Fisher Scientific, Waltham, MA, USA). Apart from the collected reference samples, Standard

Reference Material® (SRM) 2392 and 2392-I from the National Institute of Standards and Technology (NIST, Gaithersburg, MD, USA) [20, 21] were obtained. Of those, SRM® 2392 Component #1 CHR (abbreviated as SRM-C) and SRM® 2392-I HL-60 (abbreviated as SRM-H) were used as positive controls (i.e. probative samples). To monitor the presence of contamination and to assess the level of experimental and instrument noise, negative controls were introduced in each step of the workflow: reagent blanks in DNA extraction (NC-EX), as well as in long-range PCR (NC-PCR) and in library preparation (NC-LIB).

Plan of experiments and samples used are described in Supplementary Table S1. They were designed to encompass the following studies: repeatability (Supplementary Table S1a), reproducibility (Supplementary Table S1b), mixtures study (Supplementary Table S1c), concordance MPS to MPS, as well as concordance MPS to STS (Supplementary Table S1d). Simulated mixed samples were obtained by combining two persons' buccal swab sample DNA extracts in a particular ratio (0.5%, 1.0%, 2.5%, and 5.0%; Supplementary Table S1c) prior to enrichment and library prep. In mixtures study, the sensitivity of minor contributor detection was assessed, but also repeatability, since there were three replicates for each ratio of contributors. Contamination study consisted of analysing negative controls (NCs) from all sequencing runs (including, but not limited to these studies only). The general idea was to use NCs to assess the noise level and characteristics, along with assessment of noise and errors in replicates of positive controls SRM-C and SRM-H. From this information, analysis and interpretation thresholds would be calculated, and subsequently applied to other samples included in the evaluation in order to test parameters of repeatability and reproducibility of the assay.

Target enrichment, library preparation, and sequencing

A long-range PCR approach was adopted to obtain whole mitochondrial genomes in two overlapping amplicons. Primer pairs described in [22] were used (MTL-F1, MTL-R1, MTL-F2, and MTL-R2) to produce amplicons of sizes 9.1 kbp and 11.2 kbp, with the overlap covering the entire mtDNA control region. PrimeSTAR® GXL (TaKaRa, Kusatsu, Japan) was used for long-range PCR, with the following thermal cycling conditions: 25 cycles \times [98 °C 10 s + 60 °C 15 s + 68 °C 9 min 6 s] for a 9.1-kbp fragment, and 25 cycles × [98 °C 10 s + 68 °C 10 min] for a 11.2-kbp fragment. Input into target enrichment was 1 ng of genomic DNA extract in a total reaction volume of 12.5 μ L, or 2 ng in a reaction volume of 25 µL, otherwise prepared according to manufacturer's instructions [23]. Quality of PCR products was evaluated via agarose gel electrophoresis: 1% agarose gel, with the addition of 1 µL Midori Green Advanced DNA

Stain (Nippon Genetics Europe GmbH, Düren, Germany), was run for 45 min, 80 V, in SubCell® GT system (Bio-Rad, Hercules, CA, USA). Gels were visualized via GelDoc[™] system and Image Lab[™] software (Bio-Rad), whereupon they were inspected for yield, as well as for expected band size and specificity. In case of any artefacts observed by gel electrophoresis, long-range PCR was repeated for the affected sample. PCR products were quantified with Qubit[™] dsDNA High Sensitivity kit and were then normalized in a two-step manner with ultra-filtered water and resuspension buffer (RSB, from library preparation kit) down to the final concentration of 0.2 ng/µL. Equal volumes of both mtDNA amplicons were pooled for each sample, resulting in a single tube per sample, now containing entire mtDNA in two fragments. A total amount of 1 ng of each sample was taken further for library preparation, as per protocol [22].

Libraries were prepared using Nextera® XT Library Prep Kit (Illumina®, San Diego, CA, USA) according to the manufacturer's instructions [22]. Briefly, DNA was enzymatically fragmented and tagged with adapter oligonucleotides in a single reaction (tagmentation) performed by Nextera® XT transposome. Afterwards, Index 1 (i7) and Index 2 (i5) adapters were added to the tagged DNA in limited-cycle PCR. Indexed libraries underwent bead-based purification with Agencourt AMPure XP magnetic beads (Beckman Coulter, Brea, CA, USA). Afterwards, either bead-based normalization or individual normalization was applied. In the former, libraries were normalized using LNA1/LNB1 magnetic beads solution (components provided in Nextera® XT Library Prep Kit) as described in the protocol [22], while in the latter case libraries were quantified with LabChip® DNA High Sensitivity Assay on LabChip® GX Touch HT (PerkinElmer, Waltham, MA, USA) and then individually normalized to 2-3 nM using RSB. Normalized libraries were pooled in batches of 24-48 samples per run, denatured, and diluted as described in Illumina® protocol [24], with a 5% spike-in of PhiX Sequencing Control v3 (Illumina®). Paired-end sequencing was performed on an Illumina® MiSeq FGx[™] instrument using MiSeq[®] Reagent Kit v2, standard flow cell, 300 cycles (2×151 bp).

As part of the concordance study, a separate set of libraries (48 in total) was prepared using Nextera® XT Library Prep Kit from the same PCR amplicons that were used for repeatability, reproducibility, and mixtures study. Libraries were further processed in an independent laboratory by their staff: they were quantified with Agilent High Sensitivity DNA Kit on Agilent 2100 Bioanalyzer (Agilent Technologies, Santa Clara, CA, USA) according to the manufacturer's instructions, and were subsequently normalized and pooled for sequencing on Illumina® NextSeq®500 platform following protocol as described in [25]. NextSeq®500/550 Mid Output Kit v2.5, 150 cycles, was used for paired-end sequencing (2 × 75 bp). Resulting haplotypes from both MPS platforms were compared to each other for concordance, as well as to Sangertype sequencing (STS) results generated and described previously [26].

Data analysis

On MiSeq FGx[™] instrument, software Real-Time Analysis (RTA) v.1.18.54 and MiSeq® Reporter v.2.5.1.3 (Illumina®) provided primary and secondary analyses of sequencing results, applying the "mtDNA workflow" as specified in sample sheet settings prior to each run. Quality metrics were reviewed in Illumina® Sequencing Analysis Viewer (SAV) v.1.11.1 software. FASTQ files generated by MiSeq® Reporter were extracted and uploaded to Illumina® BaseSpace® Sequence Hub online platform, where they were processed by BaseSpace® mtDNA Variant Processor v1.0.0 App [27]. The application performs adapter trimming, alignment to circular reference genome, realignment of regions with indels, removal of primer contribution from reads, variant calling, read filtering and quality scoring, and generation of output files (e.g. BAM and VCF). Of the few settings that could be user-defined in mtDNA Variant Processor, common settings that were applied to all analyses comprised a minimum base call quality score for a call = 30, and genome used for alignment = rCRS (revised Cambridge reference sequence) [28,29]. Values for analysis and interpretation thresholds (AT and IT, respectively) varied: the first stage of analysis encompassed negative and positive controls analysed at AT = 0.1%, IT = 0.1%, and minimum read count = 2 (Fig. 1). This way, all signals, both true variants and false positives (noise signals and errors), were detected and taken into consideration for the calculation of thresholds, as well as for noise level assessment and characterization. All signals detected in negative controls were treated as noise originating from reagents (DNA extraction, long-range PCR, library preparation, sequencing) and/or instrument detection. Calculated values were expressed as a number of reads (read depth, DP) and included the following: minimum (MIN), maximum (MAX), average (AV), standard deviation (SD), limit of detection (LOD), and limit of quantitation (LOQ)-applying principles similar to assessing thresholds in STR markers' analysis in capillary electrophoresis [30].

Afterwards, samples of positive control samples (SRMs) were analysed in a two-fold manner:

 Data from known variants assigned to controls' haplotypes (according to [31]) were used to calculate parameters of variant quality (known as "GQ" in genome VCF files, or "Q score" in BaseSpace mtDNA Variant Analyzer reports) and percentage of homoplasmic variant (i.e. percentage required of a base in order to classify the position as homoplasmic); 2. Signals detected from all other variants not belonging to the defined haplotypes (both identical to, or differing from, rCRS) were perused similarly as in negative controls, to estimate noise level within positive controls, as well as to calculate minimum criteria for reliable variant analysis and interpretation (read depth and percentages of minor alleles), which would eventually constitute analysis and interpretation thresholds.

Overall results were used to estimate our internal analysis thresholds in terms of the minimum read depth for a reliable variant call, percentage of allele for genotype allele (i.e. calling of a homoplasmic variant at particular position), percentage of alternative allele (for point heteroplasmy calls), and genotype quality score (GQ; in Phred scale). Thus, internal analysis thresholds (INT) consisted of several components, which all variants had to comply with in order to produce a valid call.

The second stage of analysis consisted of applying the newly calculated INT to re-analyse samples of negative and positive controls to confirm the validity of thresholds. This was followed by the final stage of analysis, in which INT were applied to analyse all other evaluation samples, wherefrom repeatability, reproducibility, and concordance were assessed. At all stages of analysis, samples were visually inspected via BaseSpace® mtDNA Variant Analyzer v1.0.0 App, which allowed review of coverage profiles and sequences, as well as export to Excel-format reports. All sample reports were manually reviewed, and final variant lists (i.e. mitochondrial haplotypes) were produced for sample comparison, in accordance with the current guidelines [9, 10]. When necessary, BAM files were reviewed in an Integrative Genomics Viewer (IGV) tool v.2.4.16 [32, 33] to resolve ambiguous calls.

Results and discussion

Quality metrics assessment

Evaluation of sequencing quality (Q) metrics is an essential step in sequencing data analysis, since it is a good indicator of what to expect regarding the quality of results. High metrics quality usually means better usage of data, therefore more abundant and reliable results. All runs in this study exhibited excellent quality, as shown in the summary of selected Q metrics parameters (Table 1). Despite the variations in cluster density (491–1062 K/mm²), which were sometimes below the optimal range for MiSeq Reagent Kit v2 chemistry according to [34, 35], runs maintained a high level of quality regarding both the percentage of clusters passing filter (PF) and percentage of bases with Q score equal or higher than 30 (% Bases \geq Q30; Phred scale). Clusters PF amounted to > 90% in all runs,

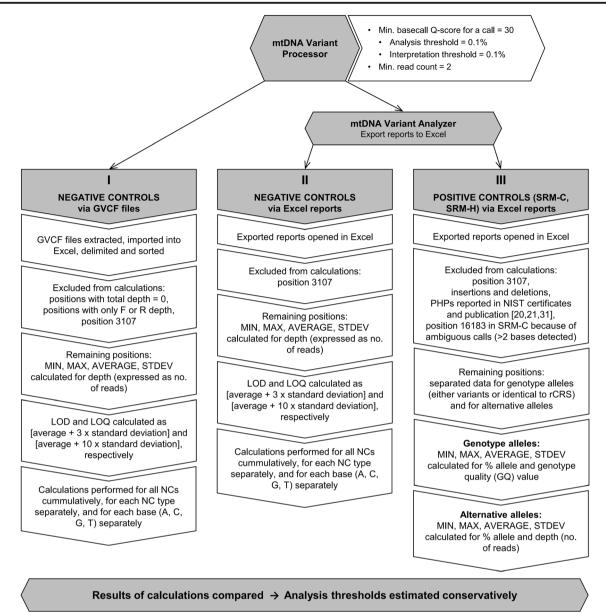


Fig. 1 Schematic diagram of analysis steps performed on samples of negative and positive controls. All controls underwent analysis in BaseSpace® mtDNA Variant Processor using identical thresholds. Genome variant call format (GVCF) files were perused in detail only for negative controls. Excel reports were perused both for negative and positive controls. After performed calculations, internal analysis, and

meaning almost all of the data were always usable for downstream analysis and, judging from % Bases \geq Q30, the great majority of bases were of sufficiently high quality for downstream analysis (variant calling, eventually). Suboptimal cluster density in runs 1, 6, and 7 affected the total yield and total number of reads PF, which in turn impacted average read depth per position per sample (Table 1) in the way that validation samples in these runs received lower average coverage than expected from calculated coverage values based on the chemistry used and targeted region (whole mtDNA). In connection to the cluster density was also the percentage of reads interpretation thresholds (INT) were defined and estimated conservatively. F, forward; R, reverse; MIN, minimum value; MAX, maximum value; AVERAGE, mean (average) value; STDEV, standard deviation; LOD, limit of detection; LOQ, limit of quantitation; NC, negative control; SRM-C, SRM® 2392 CHR; SRM-H, SRM® 2392-I HL-60; PHP, point heteroplasmy

aligned to PhiX sequencing control (% Aligned). As described earlier, we used 5% PhiX spike-in, therefore we expected % Aligned to approximate 5%. However, as spike-in percentage was in fact volume ratio, while % Aligned represented proportion of reads detected as PhiX reads in the total pool of reads PF, we observed that % Aligned in some runs deviated from the expected percentage (Table 1). Runs with high cluster density exhibited lower % Aligned and vice versa (runs with low cluster density contained more PhiX reads). Therefore, the % Aligned parameter is directly dependent on the accuracy of library quantification and subsequent loading

Run	Samples per run	Cluster density (K/mm ²)	Clusters PF (%)	Yield (gigabase)	Reads PF (million)	% Bases ≥Q30	Error rate (%)	% Aligned to PhiX	% Reads identified ^a [expected %]	Read depth a, b
1	24	491	97.8	3.0	9.4	97.1	0.5	13.1	3.9±1.8 [4.2]	5885 ± 3934
2	24	1062	91.5	5.8	18.5	93.8	0.5	4.8	4.1±1.6 [4.2]	$11,048 \pm 8683$
3	48	864	94.9	5.0	15.9	95.9	0.5	4.1	2.1±0.7 [2.1]	5084 ± 2816
4	23	939	94.1	5.3	17.0	93.0	0.5	6.5	4.0±1.3 [4.3]	9065 ± 4979
5	28	948	93.7	5.4	17.1	96.0	0.4	4.8	3.7±0.7 [3.6]	$10,935 \pm 6160$
6	30	539	97.2	3.2	10.1	96.2	0.4	11.2	3.0±0.5 [3.3]	4901 ± 2405
7	24	551	96.0	3.3	10.6	95.7	0.5	7.6	4.1±1.7 [4.2]	7770 ± 5618
8	28	745	95.8	4.4	14.0	95.3	0.6	6.1	3.4±1.2 [3.6]	8203 ± 5307

Table 1Summary of selected quality metrics parameters from evaluation runs on MiSeq FGxTM instrument. Values % Bases \geq Q30 and Error rate are given as average for the entire run

^a Expressed as average ± standard deviation; ^b only calculated for samples included in validation experiments

concentration: the former may not be as accurate using gel electrophoresis on LabChip, as opposed to qPCR [36]. The quantity of libraries may easily be over- or underestimated, thus influencing both their and PhiX's share in the total reads available (which is ultimately reflected in % Aligned value). As overclustering poses a risk to the overall success of a sequencing run, we aimed for loading concentrations safely within the manufacturer's specifications (ranging 8–15 pM) in order to avoid potential loss of quality. Judging by almost all Q metrics parameters, runs 2 and 5 displayed optimal values for our data, although the other runs were only affected in the sense of the quantity of results and not the quality, which was still well above the specifications.

Depending on the number of samples multiplexed per sequencing run, there is an expected proportion of reads identified for each library (e.g. if there are 24 samples in a run, the expected percentage of reads identified is 100/24 = 4.2% of reads assigned to each library, under condition of ideally even distribution). The values designate proportions of unique index combinations detected in the total amount of reads, and their distribution within runs gives valuable information on the efficiency of the particular lab's workflow. In our runs, percentage of reads identified for validation samples closely approximated the expected values (Table 1). Greater standard deviation was usually observed in runs where bead-based normalization was applied (runs 1, 2, and 4), as opposed to standard normalization applied in the remaining runs. It has been noted previously that bead-based normalization introduces greater variation between libraries [16].

A drop in quality was generally observed in the second read of paired-end sequencing when compared to read 1, manifesting in parameters of % Bases \geq Q30, phasing, prephasing, and error rate (Supplementary Table S2). It is not an uncommon observation, particularly since it is known that in paired-end sequencing the quality drops both in the second read, as well as towards the end of both reads [15, 37–39].

Nevertheless, this did not affect the overall quality of sequencing runs, which was unquestionable.

Regarding the coverage of mtDNA, there was a reproducible pattern across all samples: reads were unevenly distributed along the entire mitochondrial genome, with extreme drops in coverage at certain positions (Supplementary Fig. S1), regardless of sample origin (type, person, etc.). This phenomenon has been reported on numerous occasions [2, 14–16, 40], all including Nextera XT library preparation. Some readdepleted regions correspond to low-complexity (homopolymer) stretches that are known as problematic for both sequencing and alignment (e.g. positions 300-600 which harbour hypervariable segments II and III). However, the cause of coverage drops in other regions (e.g. positions 3400-3700, 5400-5600, 10,900-11,000, 13,000-13,100, 13,600-13,800) is still unknown. Some proposed that nonuniform coverage was a by-product of alignment issues because of the circular reference genome (which was shown not to be the case, after all) [14, 15], and others that it was the result of the combination of library preparation and challenging alignment [2, 16]. Still, others hypothesized that such coverage pattern resulted from Nextera XT transposome bias [16], i.e. the enzyme probably preferring certain regions of mtDNA, rather than acting randomly. We are inclined towards the latter explanation, since we observed almost identical coverage profiles in our libraries sequenced on NextSeq (data not shown) as part of concordance study, and also because it was shown that other library preparation chemistry (for example, [40]) produced different, more uniform coverage pattern. Depending on the purpose, some studies will certainly require different library preparation approach to achieve the necessary coverage uniformity—for example, uneven coverage may be acceptable for population studies (which aim for genotype variants), but less so for minor allele detection (where sufficient read depth is of paramount importance, and nonuniformity risks the loss of true variant signal).

Contamination study and noise level assessment

Library preparation protocols consist of multiple handling steps, which increase susceptibility to the introduction of exogenous contaminant DNA, facilitate cross-contamination between samples, and (by means of bead-based purification and normalization) may inflate the amount of eventual contamination-because of this, some proportion of reads is commonly found (even expected, it might be said) in NCs [2, 16, 17]. Therefore, it is recommended that NCs be introduced in various stages during library preparation, to monitor the level of background noise and the presence of contamination, so that both can be appropriately characterized and the level of tolerance established-the level below which detected noise/ contamination has no effect on results and can be classified as acceptable [2, 9, 10, 16]. To thoroughly assess the level of noise and its contents, as well as to estimate safe thresholds for reliable data analysis and interpretation, we analysed the total of 35 negative controls (NCs), sequenced as part of both assessment and other studies carried out on our MiSeq FGx instrument. Of these, 25 negative controls were reagent blanks introduced in the step of DNA extraction (NC-EX), six were amplification negative controls from long-range PCR (NC-PCR), and the remaining four negative controls were reagent blanks introduced in the step of limited-cycle PCR (NC-LIB).

In sequencing pools, NCs were represented with 0.0004– 0.0096% of the total number of reads PF. Detailed analysis of genome VCF files (GVCF) exported from BaseSpace mtDNA Variant Processor (workflow I in Fig. 1) produced the following results. Signals were detected in a total of 206,856 positions in all 35 NCs, averaging 5910 positions per NC covered with both forward and reverse reads. However, the vast majority of these positions (142,395 in total) were detected in NC-EX, out of which 91% (i.e. 129,393 positions) had a read depth of \leq 10 reads, while only 47 positions exhibited an elevated read count of > 200 reads. NC-PCR and NC-LIB consisted of a similar amount of positions with signals detected (33,699 and 30,762 respectively).

Analyses and calculations were performed both cumulatively for all NCs, for each NC-type separately, and also for each base (A, C, G, and T) to investigate potential influence of NC-type or particular dye channel (base detection) on the level and/or nature of noise signals. As shown in Table 2a, maximum depth (DP) for any NC-EX equalled 1221 reads, which is extremely high, while maximum DP for NC-PCR and NC-LIB was 57 and 21 reads, respectively. By reviewing positions with extreme read DP, we identified two regions of interest (Fig. 2): 1873–1893 (coding region, 16S rRNA) and 16128–16455 (control region, HVS-I). Region 1873–1893 showed conspicuous read depth in seven NC-EXs (> 1000 reads in one, 100–1000 reads in one, 10–100 reads in five), and in one NC-PCR (40–60 reads). Start and end coordinates of this region correlated to MTL-R1 primer, used in longrange PCR for amplification of mtDNA fragment 9.1 kbp. By visualizing BAM files in the IGV tool, we confirmed that indeed increased read depth originated from the primer (Supplementary Fig. S2, upper and middle panels). The purification of libraries may not have always been equally efficient, depending on the analyst and on handling the magnetic beads, thus a certain amount of primer might have persisted through to the sequencing. However, since we detected no signal from any of the other three primers in negative controls, it is possible this feature is specific to MTL-R1 alone. The discovery of primer signal was quite surprising, considering that all primer read contributions should have been removed by mtDNA Variant Processor [27]. For comparison, no primer reads were present in BAM files extracted from MiSeq Reporter software (Supplementary Fig. S2, lower panel), which indicates that BaseSpace application's pipeline may have issues with recognizing and removing this particular primer. Because of this phenomenon, variants detected in mtDNA positions 1873-1893 must be interpreted with caution, particularly in the case of heteroplasmy calls, since the minor allele signal might in fact originate from primer reads, instead of a true positive variant call from the sample. Most of the time, such ambiguities can be successfully resolved by visual inspection in genome browsers such as IGV. The second detected region (16128-16455), unlike the previous, was not connected to any of the primers used in long-range PCR, but it was found in eight NC-EX (> 100 reads in one, 20-100 reads in others). The presence of these two regions of increased coverage was more or less random in NC-EX and NC-PCR (independent of normalization method, analyst, number of libraries per run, etc.), and while it is an interesting observation, it also warrants caution when interpreting variant calls occurring there.

Comprehensive calculations based on all signals detected in negative controls according to workflow I (Fig. 1) were made: including primer MTL-R1 reads (Table 2a), and with primer MTL-R1 reads removed (Table 2b). Results are shown by NC-type, by base for each NC, as well as cumulative values. Following the more conservative approach, estimation of our internal analytical threshold of read depth (INT-DP) was based on the highest LOQ value. In the case when primer reads were excluded (Table 2b), estimated INT-DP equalled 100 reads. However, since primer reads could not be ignored in the analysis pipeline used, we decided to keep calculations from Table 2a, and estimated INT-DP accordingly: highest LOQ value was found in NC-EX for base G (216 reads), and by estimating the threshold at 220 reads, all signals in negative controls would have been eliminated except for the primer reads (Fig. 2). This actually corresponded well to calculations in Table 2b, because maximum read depth equalled 216 reads in any negative control after primer contribution was removed. Thus, the INT-DP threshold of 220 reads was applicable to both scenarios.

ACGT(a) MIN 2222MAX1221119712051196AVERAGE6666ST.DEV.17132113LOD57456945LOQ176136216136Read depth threshold estimation=220 rd(b) MIN 222MAX213212205216AVERAGE6666ST.DEV.9977LOD33332727LOQ96967676Read depth threshold estimation==100 rd(c) MIN 222MAX17215555163AVERAGE81058ST.DEV.1923619		NC	-PCR				NC	-LIB				All N	Cs			
MIN 2 2 2 2 MAX 1221 1197 1205 1196 AVERAGE 6 6 6 6 ST.DEV. 17 13 21 13 LOD 57 45 69 45 LOQ 176 136 216 136 Read depth threshold estimation=220 re (b)	Cum.	A	С	G	Т	Cum.	A	С	G	Т	Cum.	А	С	G	Т	Cum.
MAX 1221 1197 1205 1196 AVERAGE 6 6 6 6 ST.DEV. 17 13 21 13 LOD 57 45 69 45 LOQ 176 136 216 136 Read depth threshold estimation===20 rd (b) MIN 2 2 2 MAX 213 212 205 216 AVERAGE 6 6 6 6 ST.DEV. 9 9 7 7 LOQ 33 33 27 27 LOQ 96 96 76 76 Read depth threshold estimation==tore (c) 33 33 27 27 LOQ 96 96 76 76 Read depth threshold estimation=tore 16 76 (c) 172 155 55 163 AVERAGE 8 10 5 8 AVERAGE 8																
AVERAGE 6 6 6 6 ST.DEV. 17 13 21 13 LOD 57 45 69 45 LOQ 176 136 216 136 Read depth threshold estimation=220 rd (b) stimution=220 rd MIN 2 2 2 MAX 213 212 205 216 AVERAGE 6 6 6 6 ST.DEV. 9 9 7 7 LOQ 33 33 27 27 LOQ 96 96 76 76 Read depth threshold estimution==100 rd rd (c) stimution 2 2 2 MAX 172 155 55 163 AVERAGE 8 10 5 8 ST.DEV. 19 23 6 19	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2
ST.DEV. 17 13 21 13 LOD 57 45 69 45 LOQ 176 136 216 136 Read depth threshold estimation=220 regime (b) MIN 2 2 2 MAX 213 212 205 216 AVERAGE 6 6 6 6 ST.DEV. 9 9 7 7 LOD 33 33 27 27 LOQ 96 96 76 76 Read depth threshold estimation=100 regime regime regime regime (c) III 2 2 2 2 MAX 172 155 55 163 AVERAGE 8 10 5 8 ST.DEV. 19 23 6 19	1221	56	57	56	51	57	20	21	18	20	21	1221	1197	1205	1196	1221
LOD 57 45 69 45 LOQ 176 136 216 136 Read depth threshold estimation=220 re (b) MIN 2 2 2 2 2 MAX 213 212 205 216 AVERAGE 6 6 6 6 ST.DEV. 9 9 7 7 LOD 33 33 27 27 LOQ 96 96 76 76 Read depth threshold estimation=100 re (c) MIN 2 2 2 2 2 MAX 172 155 55 163 AVERAGE 8 10 5 8 ST.DEV. 19 23 6 19	6	5	5	5	5	5	4	4	4	4	4	6	6	6	5	6
LOQ 176 136 216 136 Read depth threshold estimation=220 rd MIN 2 2 2 MIN 2 2 2 2 MAX 213 212 205 216 AVERAGE 6 6 6 6 ST.DEV. 9 9 7 7 LOQ 33 33 27 27 LOQ 96 96 76 76 Read depth threshold estimation=100 rd (c) 172 155 55 163 AVERAGE 8 10 5 8 8 ST.DEV. 19 23 6 19	15	3	3	3	3	3	2	2	2	2	2	14	11	17	11	13
(b) MIN 2 2 2 MAX 213 212 205 216 AVERAGE 6 6 6 6 ST.DEV. 9 9 7 7 LOD 33 33 27 27 LOQ 96 96 76 76 Read depth threshold estimation=100 rd rd rd (c) IIIN 2 2 2 MAX 172 155 55 163 AVERAGE 8 10 5 8 ST.DEV. 19 23 6 19	51	14	14	14	14	14	10	10	10	10	10	48	39	57	38	45
(b) MIN 2 2 2 2 2 MAX 213 212 205 216 AVERAGE 6 6 6 6 ST.DEV. 9 9 7 7 LOD 33 33 27 27 LOQ 96 96 76 76 Read depth threshold estimation=100 re (c) MIN 2 2 2 2 2 MAX 172 155 55 163 AVERAGE 8 10 5 8 ST.DEV. 19 23 6 19	156	35	35	35	35	35	24	24	24	24	24	146	116	176	115	136
MIN 2 2 2 2 MAX 213 212 205 216 AVERAGE 6 6 6 6 AVERAGE 6 6 6 6 ST.DEV. 9 9 7 7 LOD 33 33 27 27 LOQ 96 96 76 76 Read depth threshold estimation=100 ref ref (c) (c) 172 155 55 163 AVERAGE 8 10 5 8 ST.DEV. 19 23 6 19	ads															
MAX213212205216AVERAGE66666ST.DEV.9977LOD33332727LOQ96967676 Read depth threshold estimation=100 rec (c)MIN222MAX17215555163AVERAGE81058ST.DEV.1923619																
AVERAGE 6 6 6 6 ST.DEV. 9 9 7 7 LOD 33 33 27 27 LOQ 96 96 76 76 Read depth threshold estimation=100 realized depth (c) 172 155 55 163 AVERAGE 8 10 5 8 ST.DEV. 19 23 6 19	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2
ST.DEV. 9 9 7 7 LOD 33 33 27 27 LOQ 96 96 76 76 Read depth threshold estimation=100 re (c) MIN 2 2 2 MAX 172 155 55 163 AVERAGE 8 10 5 8 ST.DEV. 19 23 6 19	216	18	19	18	19	19	20	21	18	20	21	213	212	205	216	216
LOD 33 33 27 27 LOQ 96 96 76 76 Read depth threshold estimation=100 re (c) MIN 2 2 2 2 2 MAX 172 155 55 163 AVERAGE 8 10 5 8 ST.DEV. 19 23 6 19	6	5	5	5	5	5	4	4	4	4	4	5	5	5	5	5
LOQ 96 96 76 76 Read depth threshold estimation=100 re (c) MIN 2 2 2 2 2 MAX 172 155 55 163 AVERAGE 8 10 5 8 ST.DEV. 19 23 6 19	8	3	3	3	3	3	2	2	2	2	2	7	7	6	6	7
Read depth threshold estimation=100 rd (c) MIN 2 2 2 MAX 172 155 55 163 AVERAGE 8 10 5 8 ST.DEV. 19 23 6 19	30	14	14	14	14	14	10	10	10	10	10	26	26	23	23	26
(c) MIN 2 2 2 2 2 MAX 172 155 55 163 AVERAGE 8 10 5 8 ST.DEV. 19 23 6 19	86	35	35	35	35	35	24	24	24	24	24	75	75	65	65	75
MIN2222MAX17215555163AVERAGE81058ST.DEV.1923619	ads															
MAX17215555163AVERAGE81058ST.DEV.1923619																
AVERAGE 8 10 5 8 ST.DEV. 19 23 6 19	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2	2
ST.DEV. 19 23 6 19	182	10	12	11	11	15	11	7	7	7	13	172	155	55	163	182
	12	4	5	5	5	6	4	4	4	4	5	6	9	5	7	10
	23	2	3	3	3	3	2	2	2	2	3	16	21	5	16	20
LOD 65 79 23 65	81	10	14	14	14	15	10	10	10	10	14	54	72	20	55	70
LOQ 198 240 65 198	242	24	35	35	35	36	24	24	24	24	35	166	219	55	167	210

 Table 2
 Calculations and estimations of read depth analysis threshold, based on negative control data from GVCF files (a), GVCF files excluding primer MTL-R1 reads (b), and BaseSpace® mtDNA Variant Analyzer report files (c). Negative controls originated from various stages of the workflow: DNA extraction (NC-EX), mtDNA enrichment (NC-PCR), and library

preparation (NC-LIB). Values were calculated for each base separately and then cumulatively, by each NC-type and also jointly for all NCs. Limit of detection (LOD) value was calculated as follows: average $+ 3 \times$ standard deviation. Limit of quantitation (LOQ) value was calculated as follows: average $+ 10 \times$ standard deviation

To test the validity of the estimated INT-DP threshold, calculations analogous to those described above were performed on Excel reports data exported from BaseSpace mtDNA Variant Analyzer (workflow II in Fig. 1). Reports produce lists of variants, i.e. differences from rCRS. Therefore, a large quantity of signals that are visible in GVCF files are actually not present in Excel reports, which includes primer reads (as their sequence is identical to rCRS). Nevertheless, reports may be more relevant for consideration, since a negative control (despite some portion of reads regularly expected) should not produce any variants, and no variant calls must be present in reports for NCs when validated analysis threshold is applied. Calculations resulting from negative controls' BaseSpace reports data (Table 2c) differ from the results in Table 2a and Table 2b, particularly regarding NC-EX, where greater variation among bases is evident: larger standard deviation led to higher LOD and LOQ values (highest LOQ = 242 reads for cumulative NC-EX, and LOQ = 240 reads for C in

NC-EX), even though maximum read depth detected in any NC cumulatively equalled "only" 182 reads. Estimation of read depth threshold at 240 reads, while not considerably higher than 220 reads, would nevertheless be over-conservative, since by applying the latter threshold all signals from negative controls' reports could easily be eliminated, thus establishing the tolerable level of noise below which NCs would be regarded as truly negative.

Considering the content of noise signals, i.e. whether any of the bases (A, C, G, or T) occurred more often than the other, the occurrence of each base was counted from GVCF files (Supplementary Table S3). Bases A and C were most commonly detected and in almost equal measures, followed by T, while G was the least commonly detected signal in NCs. This trend was evident in all negative controls, regardless of the type. However, since the ratio of each base count to the total number of detected bases (both by NC-type and cumulative) closely approximated its corresponding ratio in mtDNA

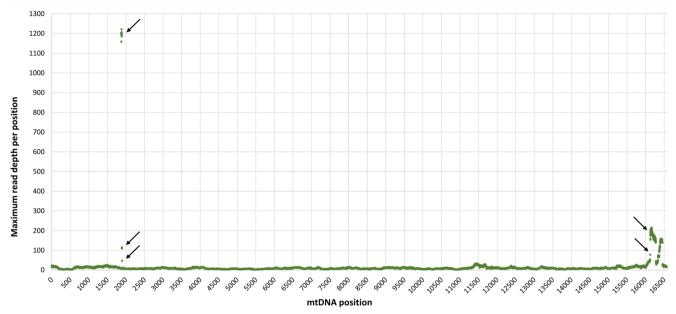


Fig. 2 Maximum read depth per mtDNA position of all signals detected in negative controls. Two regions of interest (i.e. with conspicuously high read coverage) are marked with arrows: primer MTL-R1 coordinates (1873–1893) and part of hypervariable region HVS-I (16128–16455)

(specifically, in rCRS: A = 31%, C = 31%, G = 13%, T = 25%; Supplementary Table S3), we concluded that the distribution of noise was random across the entire mtDNA, not preferring any particular base to the other.

To finalize negative control assessment, we decided to maintain the estimated read depth threshold (INT-DP) at 220 reads, which was applicable both in GVCF files analysis and BaseSpace mtDNA Variant Analyzer reports analysis. The meaning of this threshold was to set a limit for safe interpretation in terms of read depth: above the set threshold reliable variant calls can be made, and below is the area of background noise, contamination, and possible erroneous calls. Of course, detection of false positive signals is always a possibility, but the aim is to minimize that risk with carefully set thresholds, while at the same time balancing against the loss of true positive signals in the process.

Positive controls assessment

For further threshold calculations, samples of positive controls—SRM-C and SRM-H, with known and previously well-characterized sequences [20, 21, 31]—were analysed according to workflow III in Fig. 1. Indels in hypervariable regions (HVS I–III) were ignored at this time, along with point heteroplasmy variants that were reported in [31], since they cannot be considered as either errors or true variant calls until validated thresholds are applied. Therefore, indels and point heteroplasmy (PHP) calls were excluded from calculations, as well as position 16183 in sample SRM-C. The latter was detected as an ambiguous variant call: a mixture of two bases (A and C) and deletion. It is in fact homoplasmic variant A16183C which, in conjunction with T16189C (also present in SRM-C), produces uninterrupted homopolymer stretch of 11 cytosines, resulting in alignment issues, which were reported and elaborated in [31]. As mentioned earlier, ambiguities such as this may be resolved in most cases by visual inspection of read alignment in tools such as IGV.

In eight samples (four replicates of each SRM), signals were detected in 3280 positions, in total. Single bases (100% variant from rCRS) were called for 194 positions, while in all other positions between one and three alternative alleles were detected (bases and/or deletions) in addition to the major base. Calculations were performed cumulatively for all SRMs, and also separately for each base (Table 3).

Regarding read depth calculations for alternative alleles (Table 3a), results were concordant with those obtained for negative controls, wherefrom estimated coverage threshold (INT-DP) of 220 reads would be applicable to SRMs as well. Although cumulative LOQ was considerably lower (125 reads), we decided to keep the minimum read count at 220 reads, since the highest LOQ was calculated for C (207 reads; Table 3a), which is just short of the estimated negative controls' threshold (Table 2). As visible in Fig. 3a, there are two positions where the maximum read depth of the alternative allele exceeded the threshold: in particular, variants detected were A2487M and T16189d. However, these two would not be taken into consideration for true variant calls: the former exhibited extremely poor GQ value (26-29, Phred score) in both control samples, and the latter consisted of ambiguous calls (C and deletion, or C and T and deletion) only in SRM-C, mirroring the same problem described above for the A16183C-in this case, variant T16189C contributed to the prolongation of homopolymeric C-stretch and subsequent issues in alignment.

 Table 3
 Thresholds based on data from BaseSpace® mtDNA Variant

 Analyzer report files of positive controls SRM-C and SRM-H (SRM®

 2392 CHR and SRM® 2392-I HL-60, respectively). Thresholds for read

 depth and percentage of alternative alleles were calculated and estimated

 from alternative (non-haplotype) alleles detected in positive controls (a).

Thresholds for genotype (homoplasmic) alleles and quality scores (GQ; Phred scale) were calculated and estimated from known haplotype variants in positive controls (b). Limit of detection (LOD) value was calculated as follows: average + $3 \times$ standard deviation. Limit of quantitation (LOQ) value was calculated as follows: average + $10 \times$ standard deviation

	А	С	G	Т	Cum.	А	С	G	Т	Cum.
(a)	Read dep	th				Alternat	ive allele			
MIN	2	2	2	2	2	0.1%	0.1%	0.1%	0.1%	0.1%
MAX	46	449	130	155	449	1.3%	5.5%	0.6%	2.4%	5.5%
AVERAGE	3	7	4	3	5	0.1%	0.2%	0.1%	0.1%	0.2%
ST.DEV.	4	20	8	7	12	0.1%	0.5%	0.1%	0.1%	0.3%
LOD	15	67	28	24	41	0.4%	1.7%	0.3%	0.4%	0.9%
LOQ	43	207	84	73	125	0.9%	5.1%	0.6%	1.1%	2.8%
	Estimated	l read depth (threshold=21	0		Estimate	ed alternative	allele thresh	old=3%	
(b)	Genotype	allele				GQ scor	e			
MIN	94.5%	94.0%	99.4%	96.5%	94.0%	25	31	27	27	25
MAX	100.0%	100.0%	100.0%	100.0%	100.0%	50	50	49	50	50
AVERAGE	99.8%	99.9%	99.9%	99.9%	99.9%	46	48	46	47	47
ST.DEV.	0.4%	0.2%	0.1%	0.2%	0.3%	5	3	4	4	4
LOD ^a	98.6%	99.2%	99.7%	99.1%	99.0%	41 ^b	45 ^b	42 ^b	43 ^b	43 ^b
LOQ ^a	95.8%	97.6%	99.2%	97.5%	97.2%					
	Estimated	l genotype all	ele threshold	=97%		Estimate	ed GQ score (threshold=41	l	

^a Since the max. value is 100%, LOD and LOQ calculated as [average $-3 \times$ standard deviation] and [average $-10 \times$ standard deviation], respectively; ^b standard LOD and LOQ formulas not applicable (since GQ scores are in Phred scale), therefore calculated as follows: average $-1 \times$ standard deviation

The other parameter calculated from alternative allele signals was percentage of minor alleles, with the maximum of 5.5% (Table 3a, Fig. 3b) detected at A2487M—the same position that showed elevated read depth earlier. Estimating from the cumulative calculated LOQ, analytical threshold for minor alleles (INT-AN) would be 3%. By applying this threshold, 99% of signals would be successfully eliminated, since in the total of 3155 alternative alleles detected in all SRMs, only 39 were > 1%. However, as evident from Table 3a and Fig. 3b, alternative alleles with considerably higher minor allele percentages may occur, and that prompted us to establish additional, interpretation threshold for minor alleles (INT-IT) which equalled 6%. The meaning of this dual-threshold system is as follows: PHP calls with alternative (minor) alleles > 6% are safe for interpretation, under condition of sufficient read depth; PHPs with minor alleles between 3% and 6% are required to undergo additional scrutiny of other quality parameters before they are reported; minor alleles < 3% are in the area where it is virtually impossible to distinguish between noise signals and true positive calls (without alternative confirmation method), therefore they cannot be reported as such.

Regarding variants reported as haplotypes (i.e. genotype alleles, GT), calculations were performed analogously to the ones described for alternative alleles above (Table 3b). As a result, the threshold for homoplasmic genotype alleles (INT-GT) was estimated at 97% according to cumulative calculations. Notably, minimum values detected for bases A and C were < 97% (94.5% and 94.0%, respectively; Table 3b), but by additional review, we found that the minimum signal for A originated from A2487M, a low-quality variant call, while the minimum for C was in fact caused by the sum of two minor alleles at the same position (namely, 2.4% T and 3.6% deletion). Overall, we decided to keep the estimated genotype variant threshold (INT-GT) at 97%, meaning that at any position a variant allele exceeding 97% would be considered homoplasmic, i.e. single-base variant call—no PHP call would be allowed for this position. This is in accordance with previously calculated minor allele analysis threshold (INT-AN) of 3%.

In addition to the threshold of percentages for genotype alleles, we performed calculations for quality values (GQ) of genotype positions (Table 3b). Since the use of standard LOD and LOQ formulas (i.e. $3 \times$ and $10 \times$ standard deviations from average, respectively) was not feasible in this case, we opted for a modified formula more appropriate for the GQ values: average $-1 \times$ standard deviation. The cumulative GQ threshold (INT-GQ) equalled 43 (Table 3b); however, we decided to keep the threshold at 41 to accommodate for values of all bases (and calculations for base A produced the value of 41). Intriguingly, position 2706 exhibited GQ lower than other genotype positions in SRMs (GQ 37–41), but also in all other analysed samples (GQ values ranging from 33 to 49, of which more than 80% were < 41). Because of this, and similar

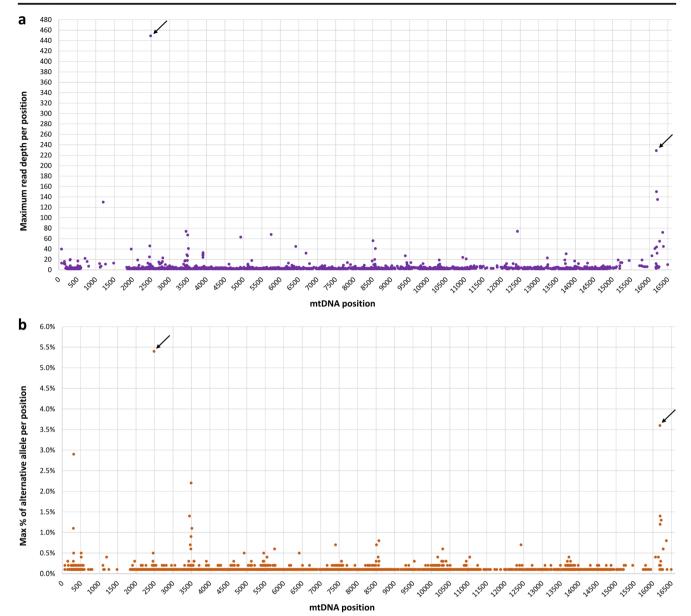


Fig. 3 Graphical representation of maximum read depth of alternative alleles per position (a) and maximum percentage (%) of alternative alleles per position (b) in positive control samples (SRM® 2392 CHR

exceptions to the other threshold components, we must bear in mind that, for a reliable variant call, thresholds defined for all parameters must be met and considered as a whole, rather than as individual, independent requirements.

Finalized definition of analysis thresholds

Based on the calculations described in previous sections, we finalized the values proposed as our internally evaluated thresholds (INT) for whole mtDNA analysis in high-quality samples, encompassing multiple parameters:

INT-DP = 220 reads

and SRM $\$ 2392-I HL-60). Extremes detected in positions 2487 and 16189 (on both graphs) are marked with arrows

- INT-GT = 97%
- INT-GQ = 41
- INT-AN = 3%
- INT-IT = 6%

Accordingly, we defined our internal guidelines for whole mtDNA analysis and interpretation as follows:

- A minimum depth of 220 reads is required for a variant allele to be taken for analysis.
- Quality score (GQ) ≥41 is required for a position to be reliable for variant calling. Otherwise, the position is most likely to contain erroneous variant calls.

- All positions with a major allele $\ge 97\%$ are considered homoplasmic and a single-base variant is called.
- Alternative alleles < 3% are not analysed nor interpreted, since they reside within the area of background noise.
- Alternative alleles between 3% and 6% are taken into analysis. They may be interpreted and subsequently reported, if read depth and quality score thresholds are complied with.
- Alternative alleles ≥6% are considered safe to interpret and report, since presumably, all other thresholds' criteria have already been fulfilled.

At first glance, the read depth threshold of minimum 220 reads may seem overly conservative, but its greatest advantage is that it was derived from our own experimental data, rather than set arbitrarily or taken at set value from other studies (e.g. [15–17, 40]). Detection of minor allele present at 3% would hereby require a depth of 7333 reads, while detection of minor allele at 6% would require a depth of 3667 reads. Despite large read counts, these requirements are easily met, since multiplexing of 24 samples per run gives theoretical coverage of 9375 reads per position per sample. Even multiplexing as many as 48 samples per sequencing run gives theoretical coverage of 4688 reads per position per sample, which is ample enough for detection of minor alleles with frequencies of 4.7% and higher. The only obstacle to detection of minor alleles is uneven coverage across the mitochondrial genome, which displays some chemistry- and sequencedependent profile, as described earlier. Therefore, detection and interpretation of minor allele signals in presumably heteroplasmic positions should be mindful of shortcomings specific to the method used.

In addition to our internal guidelines elaborated above, interpretation and calling of indels should not be based solely on percentages obtained from BaseSpace mtDNA Variant Analyzer reports. Read alignments for any indel call are to be manually inspected by visualization in genome browsers such as IGV, prior to determining the dominant molecule [9], which would be reported as the final variant call.

Here we presented our approach to the calculation of analysis thresholds, which uses a multiple-parameter system to define internal guidelines for analysis and interpretation of whole mtDNA MPS results (something similar has been done in [41] for interpretation of negative controls). As the studies were performed in a forensic laboratory, the aim was to maintain similarity to the method traditionally used to derive thresholds in forensic STR markers' analysis via capillary electrophoresis. As prescribed by [10], each laboratory should develop and implement their individual interpretation guidelines based on validation and evaluation studies, which is what we aimed to do here for our own data. This approach is applicable for other laboratories performing similar studies, but it is possible that the actual threshold values would slightly vary, since each laboratory presents a unique system with its staff, equipment, consumables, and environment. It is also important to note that samples used herein consisted of control samples and high-quality reference samples. Therefore, while the principles of setting the thresholds are applicable to other forensic samples and methodologies, analysts should be wary as challenging forensic samples requires a different library preparation approach and may subsequently warrant the introduction of additional analysis and interpretation guidelines into laboratory workflow.

Repeatability

Definition of repeatability in general terms, according to [7, 8], is the variation in measurements of results obtained by the same person (analyst) multiple times on the same instrument. This can be applied two-fold to the sequencing library preparation workflow, since replicates of a sample may consist of PCR replicates (same sample amplified in multiple PCR reactions and from each a separate library prepared) and library replicates (i.e. technical replicates, meaning multiple libraries prepared from the same PCR reaction of a sample). Having that in mind, we tested repeatability by comparing final variant calls (final haplotypes) of PCR replicates and library replicates for the samples of buccal swabs and blood on FTATM Cards ("B" and "F", respectively) of persons MW-0002 and MW-0020 (schedule in Supplementary Table S1a). Final haplotypes from library replicates of positive controls SRM-C and SRM-H were compared for repeatability as well. In all instances, indel and heteroplasmy calls underwent additional review and visual confirmation of read alignment in IGV. Repeatability was assessed for two analysts separately, to evaluate the variation of library preparation between different persons handling the protocol.

Library replicates of sample MW-0002-B showed 100% repeatability, regarding final variant calls, for both Analyst 1 and Analyst 2. PCR replicates of MW-0002-B showed complete repeatability as well, regardless of the analyst. Both library and PCR replicates of sample MW-0002-F exhibited 100% repeatability, including point heteroplasmy T16311Y, which was consistently called across all replicates (Supplementary Table S4). In most replicates of sample MW-0020-B, there were two PHPs consistently detected: T152Y and T9325Y (Supplementary Table S4). The few exceptions occurred in instances where read depth of the minor allele did not exceed the required threshold of 220 reads, and thus required manual review below the validated thresholds. In these cases (8 in total; Supplementary Table S4a and S4b), were it not for multiple replicates for comparison, these calls would pass as homoplasmic variants. However, for the purpose of this study, the presence of minor allele was considered confirmed, even for those with fewer reads than necessary. For the sample MW-0020-F, only library replicates were made, and they exhibited complete repeatability. One PHP was detected, T9325Y, which was consistently called in all replicates (Supplementary Table S4b).

Regarding technical replicates of positive controls, SRM-C exhibited 100% repeatability, including one PHP position (C64Y), which was consistently detected in all three replicates, and is concordant with [31]. Haplotypes of SRM-H replicates were repeatable as well, altogether with three heteroplasmy calls: T2445Y, C5149Y, and T12071Y. Percentages of minor alleles detected were in accordance with [31] for all three PHPs. However, only T12071Y was completely repeatable (most likely due to a larger proportion of minor allele), whereas for both other PHPs one or more deficiencies were observed. Read depth requirement was not met in one of three replicates for both T2445Y and C5149Y, and manual review was necessary to confirm the presence of minor allele. Besides read depth, heteroplasmy T2445Y proved more complex to interpret after application of our validated thresholds, since in all replicates GQ fell below 41 (Supplementary Table S4a) for this position. Upon inspection of this variant's environment, we determined that it resides within a region where a drop in GQ is prominent, in all replicates, and encompasses positions 2412-2487. Thus, we recommend that any variants be interpreted with caution, as this region is obviously prone to quality issues in general (the same phenomenon was observed across all samples and sample types). Regarding T2445Y in SRM-H, since it was detected in all replicates and was described previously [31]-even though the question of quality was not discussed there-this heteroplasmy was reported and included in repeatability assessment in this study. Were it not for multiple replicates and literature confirmation, the T2445Y variant would likely be omitted from the final haplotype due to not meeting all threshold criteria.

Overall, 783 variant calls (differences from rCRS) were reviewed in the course of the repeatability test, across 43 replicates in total. For Analyst 1, 564 variant calls were assessed in total, out of which six calls were discrepant (1.1%). Similarly, in the case of Analyst 2, out of 219 variant calls that were assessed in total, two of them showed discrepancy (0.9%). Thus, repeatability equalled 98.9% and 99.1% for Analysts 1 and 2, respectively. Since discrepant calls exclusively concerned point heteroplasmies, whereby manual review confirmed the presence of minor alleles, the whole assay was appraised as completely repeatable.

Reproducibility

Reproducibility study encompassed comparison of haplotypes for two sample types of 11 persons, along with positive controls SRM-C and SRM-H. Analyst 1 and Analyst 2 independently prepared batches of libraries, which were sequenced in separate runs. As previously described for the repeatability study, final variant calls (haplotypes) of samples were compared, while indels and heteroplasmy calls required additional confirmation in the IGV tool to be considered for comparison.

Out of 26 pairs of haplotypes that were compared in total, six exhibited some form of discordance and were manually reviewed to determine the cause. In all cases, the main reason for observed discrepancies was inconsistently called PHPs in one sample of the pair (Supplementary Table S5). Samples MW-0078-B, MW-0020-B2, MW-0065-F, MW-0067-F, and SRM-H all exhibited heteroplasmies detected in the results of Analyst 2, while apparently no corresponding heteroplasmy call was found in the results of Analyst 1. The presence of minor alleles, as described in the previous section, was established by manual review below the validated thresholds (220 reads), and in all instances, heteroplasmy calls were confirmed. For the purpose of this study, such results were considered reproducible.

While the same effect was observed in sample MW-0087-B (variant T8955Y was detected only in one of the pair, and seemingly no minor allele signal, i.e. 0%, was detected in the other), the cause was different. To resolve this, we lowered the analysis threshold below 3%, and found minor allele C at 2.9%, despite excellent read depth (396 reads; Supplementary Table S5). Thus, heteroplasmy call was considered confirmed for the purpose of reproducibility, even though normally it would not be detected as PHP since it does not comply with all components of our validated thresholds.

Additionally, to serve as our own internal control sample, MW-0020-B was sequenced in all our runs, 18 times in total (not limited to evaluation runs only). These results were included as part of the reproducibility study, since they encompassed five different analysts who prepared libraries, and multiple runs. Haplotypes were fully reproducible, regardless of analyst and run, including two PHP calls, T152Y and T9325Y (Supplementary Table S6). Percentages of minor alleles were consistent with results from Supplementary Table S4 and Supplementary Table S5. Along with quality (GQ) and read depth (DP) parameters, they confirm the validity of our "dual" threshold system for analysis and interpretation, since all PHPs between 3 and 6% of minor allele conform to other INT components (GQ and read DP; Supplementary Table S6), and are therefore safe to interpret and report (after analyst review) according to our validated thresholds.

Overall, the assay produced reproducible results between analysts and different runs. The exceptions were few cases of inconsistent heteroplasmy calls: of 724 pairs of variants compared for reproducibility in total, seven pairs required manual analyst review as one of the pair did not meet a component of thresholds' criteria. Nonetheless, heteroplasmy calls were eventually confirmed, and thus considered reproducible as well in this study.

Concordance

Concordance study consisted of two parts: firstly, MPSgenerated mtDNA haplotypes were compared to STS results (published previously as part of Croatian population study [26]); and secondly, MiSeq-generated results were compared to NextSeq-generated results, obtained by the same library preparation reagents, but sequenced in an independent laboratory on a different instrument.

MPS to STS

We compared haplotypes of 10 persons' buccal swabs used in this study to their corresponding haplotypes generated by STS. The latter encompassed only the mtDNA control region, while in this study we sequenced whole mtDNA. In general, results were concordant (Supplementary Table S7), with few exceptions concerning PHP calls, as well as insertions. For example, insertions at position 573 were regularly detected in ranges of 3-10% (as reported in Excel reports from BaseSpace mtDNA Variant Analyzer application), which is far below the 50% required to call the dominant molecule. However, these percentages may not reflect the actual state: they may have been artificially produced (or, rather, reduced) by alignment artefacts. Therefore, by viewing read alignments via the IGV tool, we were able to resolve apparent discrepancies between STS and MPS: insertions 573.1C-573.3C were confirmed in MW-0012, insertions 573.1C-573.4C confirmed in MW-0026 and MW-0067, insertion 16193.1C confirmed in MW-0065, and insertions 16193.1C-16193.2C confirmed in MW-0078. The presence of insertions was sufficient to appraise results as concordant, since length variation cannot be counted as exclusion [9, 10], or discordance in this case.

Apart from indel calls, which were manually reviewed and confirmed, point heteroplasmies were the main source of discrepancies, as expected, since MPS readily detects minor alleles below 10%, which is the nominal sensitivity of detection for the STS method. Thus, samples MW-0020, MW-0067, MW-0087, and MW-0088 exhibited PHPs that were not seen previously in STS results: T152Y, C16301Y, A374R, and C16256Y, respectively (Supplementary Table S7). These observations were not unexpected, since in all four PHPs minor allele proportions were < 10% (Supplementary Table S5), and thus passed undetected by STS. Furthermore, samples MW-0026, MW-0065, and MW-0078 exhibited homoplasmic variants in STS results (T16093C, T16093C, and A200G), whereas MPS revealed these positions as actually heteroplasmic (Supplementary Table S5 and Supplementary Table S7). Minor allele T might have been detected by STS in sample MW-0026, since proportions from STS results exceeded 11%, however, the observation was probably not sufficiently confident for the PHP call.

In general, MPS-generated results were concordant with STS-generated results, with few exceptions like indels and PHP calls, the first due to MPS method limitations (bioinformatic solutions still struggle with homopolymeric nucleotide stretches and other low-complexity regions, thus creating artificial image of indels), and the latter due to STS method limitations (sensitivity of minor allele detection). Besides comparison of control region haplotypes, MPS of whole mtDNA evidently generates much more information and greatly complements STS data. It is particularly elucidating to see the number of variants arising in the coding region, as well as the appearance of more heteroplasmic positions. This gain of discriminatory information would be particularly relevant for forensic purposes.

MPS to MPS (MiSeq to NextSeq)

To validate our whole mtDNA MPS results, 36 pairs of haplotypes were compared for concordance assessment between two MPS platforms: MiSeq FGx in our laboratory and NextSeq in an independent laboratory (Supplementary Table S1d). MiSeq data were analysed at the established INT thresholds, with indels and heteroplasmy calls subsequently reviewed via the IGV tool as described previously. The exact analysis thresholds, however, could not be applied to data from NextSeq instrument-different instrument, different operators, and different laboratory environment-at least not without conducting a separate evaluation to establish thresholds specific to that instrument's conditions, which was beyond the scope of this study. Therefore, all variants detected on MiSeq and reported in final haplotypes of samples only sought confirmation in the NextSeq data, and not complete compliance with the calculated INT thresholds.

The majority of samples showed absolute concordance between results from the two sequencing platforms. Some minor discrepancies were noted, arising from heteroplasmy calls (Supplementary Table S8). For samples MW-0020-B and SRM-H, which had two and three PHPs detected, respectively, one of the three library replicates of each sample exhibited low coverage of minor alleles in MiSeq results (read depth < 220 reads; Supplementary Table S8). Normally, if that one replicate were uniquely sequenced sample either for MW-0020-B or SRM-H, MiSeq calls would not have been defined as heteroplasmies, but as single variants. However, since these particular variants were detected in all other replicates of MW-0020-B and SRM-H, multiple times during repeatability and reproducibility studies (Supplementary Tables S4-S6), here they were acknowledged as PHPs as well. The presence of minor alleles for all PHPs in those two samples was unambiguously confirmed in NextSeq results, which offered much better coverage, and subsequently easier interpretation.

Further, in all three replicates of sample MW-0020-F, variant T9325Y was underrepresented in the NextSeq data, regarding both minor allele percentage and read depth (< 3%) and < 220 reads, respectively). It is worth noting that these replicates received less than average share of reads: 0.07-0.57% reads identified, while approximately 1% would be expected since 96 samples were multiplexed for the NextSeq run. Consequently, read depth was lower in these samples, and some variants were very poorly covered (e.g. only 22 reads for minor allele C in replicate MW-0020-F2). Regardless of that, the presence of minor allele was established in all replicates and was sufficient for the confirmation of concordance. By the same analogy, heteroplasmy C16301Y in sample MW-0067-F showed minor allele at 2.9% in the NextSeq dataset, and though it may be below the established thresholds on MiSeq, it was not considered as a discordance since the confirmation was all that we needed from NextSeq.

In contrast to the reproducibility study (Supplementary Table S5), sample MW-0080-B showed additional heteroplasmy call (T16093Y). Probably it passed undetected earlier because of poor read depth and/or minor allele < 3%. However, it was now detected on MiSeq, and also confirmed in its corresponding pair mate in NextSeq results (Supplementary Table S8). Adversely, samples MW-0087-B and MW-0065-F experienced a loss of heteroplasmy call (T8955Y and T16093Y, respectively), in comparison to reproducibility study results (Supplementary Table S5), as their respective minor alleles probably lacked either read depth or percentage to be detected. These observations were not surprising for either of these samples, since all three heteroplasmies exhibited minor allele proportions on the borderline of the defined analysis thresholds for MiSeq data (very close to 3%), and thus may or may not be detected, which strongly depends on sequencing run metrics in each particular case.

Overall, comparison of sequencing results comprised a total of 955 pairs of variants (differences from rCRS) between two MPS platforms. In several instances, manual review was required before confirmation of results, but they were all successfully resolved. Both datasets unequivocally showed complete concordance, as expected, since both instruments originate from the same manufacturer, and are based on the same sequencing-by-synthesis technology.

Mixtures study

As part of the repeatability study, but also to test the reliability of minor allele detection in heteroplasmy calls, as well as to discriminate between true PHPs and contamination events (manifesting as mixtures), we prepared simulated mixed samples (Supplementary Table S1c). Buccal swab samples of two female persons MW-0002 and MW-0020 were selected, since they were previously used for repeatability studies, thus sequenced multiple times, and their sequence was by now well known. They were combined in the ratios 1:199 (MIX-1 = 0.5%), 1:99 (MIX-2 = 1.0%), 1:39 (MIX-3 = 2.5%) and 1:19 (MIX-4 = 5.0%). Mixed samples underwent long-range PCR (three replicates each) and library preparation protocol as previously described for all other validation samples. The two haplotypes differed in exactly 12 positions (4 in the control region, 8 in the coding region; Supplementary Table S7), which we targeted for analysis with the lowered thresholds. Other positions were not eligible for analysis and interpretation, since mixture ratios were mostly below the thresholds established by this evaluation.

Read depth for the targeted positions varied (minimum 1461 reads; maximum 30,102 reads), but in all instances, it was sufficient for the detection of minor contributor at the expected ratios. Minor contributor was successfully detected in all mixtures at the expected mtDNA positions. However, percentages of minor contributor alleles differed from the theoretical values: on average, in all four mixtures, minor contributor was detected in excess of the expected ratio (Table 4). It was interesting to note that at positions 2259, 4745, and 14872, minor contributor alleles were detected with as much as twice the expected ratio (e.g. 1% instead of 0.5%, 10% instead of 5%, etc.). This particular position-specific phenomenon remains inexplicable, since these mtDNA positions do not reside within error-prone regions, neither does the major contributor exhibit additional PHPs at these coordinates which would tilt the ratios to such extent. Contributing to this unusual phenomenon is the fact that NextSeq results (as mixtures were sequenced alongside other samples in concordance study) showed identical trend, and almost identical values, among minor contributor ratios, for exactly the same three positions (data not shown).

One possible explanation for the difference between average observed minor contributor ratios and expected values is that it might have been caused by bias during long-range PCR: one contributor's mtDNA might have been amplified more efficiently than the other's. This would introduce slight change to the ratio of contributors from the start and eventually it would manifest itself in the results. Alternatively, as indicated in [17], the skewed observed ratios may more likely be the product of differences in mtDNA vs. nDNA quantity between samples: in that case, expected mixture ratios calculated from genomic DNA concentrations would not exactly correspond to the final results where mtDNA to mtDNA ratios were observed. Notwithstanding, whole mtDNA workflow in general consists of multiple steps wherein ratios of contributors may be affected. Thus, even though sequencing is reproducible and relatively precise, proportions of the minor contributor in mixed samples can only be assessed approximately by this method since multiple preparation steps, in combination with the varying content of mtDNA within the sample, may introduce bias to the ratio of contributors.

rCRS position rCRS base	7 T T	2259 C F	2706 A	4745 A C	5897 C	7 028 C	1 3680 C H	14872 C T	15904 C C	16231 T T	16298 T T	16359 T T				
Minor allele	- U	- U	Q Q	D A) H		- U	- U	Ч	- U	- U	- U	Observed average ± SD	Expected	Observed ratio	Expected ratio
MIX-1-1 MIX-1-2	$0.7\% \\ 1.1\%$	$0.9\% \\ 1.0\%$	0.6% 0.7%	1.1% 1.2%	$0.5\% \\ 0.7\%$	0.6% 0.9%	$0.3\% \\ 0.7\%$	0.7% 0.9%	$0.5\% \\ 0.7\%$	0.6% 0.9% 0.9% 0.9% 0.0% 0.0% 0.0% 0.0% 0.0	0.5% 0.7%	0.6% 0.8%	$0.7 {\pm} 0.2\%$	0.5%	1:135	1:199
MIX-1-3	0.7%	0.9%	0.6%	1.2%	0.6%	0.5%	0.6%	0.9%	0.6%	0.9%	0.6%	0.6%				
MIX-2-1	1.4%	2.1%	1.2%	2.8%	1.4%	1.5%	0.7%	1.5%	1.2%	1.4%	0.9%	1.4%	$1.3 \pm 0.4\%$	1.0%	1:76	1:99
MIX-2-2	1.3%	1.3%	1.1%	2.0%	1.0%	1.3%	1.1%	1.6%	1.3%	1.5%	1.1%	1.2%				
MIX-2-3	1.1%	1.6%	1.2%	1.7%	1.1%	1.3%	0.9%	1.2%	1.0%	1.3%	0.9%	1.0%				
MIX-3-1	3.2%	4.8%	2.9%	5.3%	3.2%	3.5%	2.0%	3.1%	3.1%	3.2%	2.2%	2.6%	$3.3 \pm 0.7\%$	2.5%	1:31	1:39
MIX-3-2	3.3%	4.4%	3.3%	4.5%	3.2%	3.6%	2.4%	3.8%	3.1%	3.5%	2.5%	2.9%				
MIX-3-3	3.2%	4.0%	2.6%	4.5%	2.9%	3.1%	2.5%	3.7%	2.9%	3.2%	2.2%	2.8%				
MIX-4-1	6.6%	9.0%	5.4%	9.4%	6.1%	6.7%	4.2%	8.2%	6.3%	6.7%	5.0%	6.0%	$6.5 \pm 1.6\%$	5.0%	1:15	1:19
MIX-4-2	6.4%	8.4%	5.1%	9.8%	6.1%	6.5%	4.1%	7.2%	5.8%	6.6%	4.7%	5.6%				
MIX-4-3	6.2%	8.9%	5.5%	10.6%	6.0%	6.7%	4.4%	7.1%	5.5%	6.4%	4.5%	5.6%				

Besides detection of minor contributor, we monitored the presence of two PHPs characteristic to the buccal swab sample of MW-0020, as described in previous sections (Supplementary Tables S4-S6). Both heteroplasmies (T152Y and T9325Y) were consistently called in all mixtures (Supplementary Table S9), regardless of the proportion of minor contributor, and their respective values correspond well to the minor allele percentages reported in previous experiments of this study.

Conclusion

Based on multi-component criteria of data analysis thresholds (in terms of read depth, percentage of alleles, and quality scores), which were established in this study, we defined internal guidelines for analysis and interpretation of mtDNA results obtained by MPS. The proposed methodology proved robust and confident for variant calling and reporting when applied to analysis of controls and reference samples alike. Our study also shows that the whole mtDNA assay on MiSeq FGxTM produces repeatable and reproducible results (both between runs for the same analyst, and between different analysts) for all samples, equally for buccal swabs and blood samples, as well as for cell-culture-derived positive control samples (SRMs 2392 and 2392-I). Moreover, results were completely concordant with STS results [26] and were also concordant with results obtained on another MPS platform. Few minor discrepancies were observed, originating from heteroplasmy calls that did not comply with at least one component of defined analysis thresholds, but all calls were eventually confirmed in both datasets after analyst review; thus, no major discordance was noted. We conclude that this assay-including enrichment strategy, library preparation reagents, sequencing reagents, sequencing instrument, and accompanying analysis software-is suitable for further use in our forensic laboratory, primarily for samples of good quality, such as reference samples and/or high-quality stains. It will be further used for Croatian population study on whole mitochondrial genomes, in order to establish a national database for the purpose of haplotype and haplogroup frequencies.

Some features of the analysis software may require additional attention in future upgrades, for example, dealing with leftover primer reads, treatment of indels and homopolymeric regions (a common struggle to almost every mtDNA analysis program), accommodation of forensic mitochondrial nomenclature, and also making more parameters available for user-modification in order to better tailor the analysis to specific study goals. All in all, Illumina® BaseSpace® Sequence Hub online bioinformatics platform is, at present, an acceptable solution for fast, intuitive, high-throughput data analysis which will be required for the population study.

Free online, cloud-based platforms such as BaseSpace®, with its plethora of applications, can be user-friendly, require little previous bioinformatic knowledge, and provide simple, fast, cost-effective solutions to streamline both data analysis and data storage. However, online solutions are unsuitable in a forensic setting, where data handling procedures are strictly prescribed by laws and protocols, dedicated off-line servers are used for analysis and storage of sensitive caserelated information and analysis results in order to maintain their confidentiality, etc. Considering that, at some point in the future, whole mtDNA analysis by MPS will be implemented into routine forensic casework, the choice of analysis software will have to be reconsidered. Therefore, it is imperative that, in parallel to the population study, in the future, a comparison of other available analysis software be conducted, in order to decide the best bioinformatics solution for casework samples. Needless to say, they provide more challenge than reference samples used in evaluation and population studies, and would thus require a different approach not only in terms of analysis software, but in library preparation method as well.

Supplementary Information The online version contains supplementary material available at https://doi.org/10.1007/s00414-021-02508-z.

Acknowledgements The authors thank all participants in the study for their valuable contributions in the form of samples and detailed informed consents. The authors are also thankful to Oliver Vugrek, PhD, Head of Laboratory for Advanced Genomics, Division of Molecular Medicine at "Ruđer Bošković" Institute, and their laboratory staff for collaboration in concordance study. Theauthors are grateful to Sara Rožić and Ivana Račić, PhD, who made valuable contributions in the experimental part of this study.

Funding This work was supported by the Ministry of the Interior of the Republic of Croatia.

Data availability The datasets generated and analysed during this study are available from the corresponding author on reasonable request.

Compliance with ethical standards

Conflict of interest The authors declare that they have no conflict of interest.

Ethics approval This study involved samples collected from human participants. All procedures performed in the study were in accordance with the institutional and national ethical standards.

Consent to participate Informed consent was obtained from all individual participants included in this study.

Consent for publication Not applicable.

Code availability Not applicable.

References

- Butler JM (2012) Mitochondrial DNA analysis. In: Butler JM (ed) Advanced topics in forensic DNA typing: methodology. Academic Press, Elsevier, Cambridge, pp 405–456
- Parson W, Huber G, Moreno L, Madel MB, Brandhagen MD, Nagl S, Xavier C, Eduardoff M, Callaghan TC, Irwin JA (2015) Massively parallel sequencing of complete mitochondrial genomes from hair shaft samples. Forensic Sci Int Genet 15:8–15
- Lyons EA, Scheible MK, Sturk-Andreaggi K, Irwin JA, Just RS (2013) A high-throughput Sanger strategy for human mitochondrial genome sequencing. BMC Genomics 14:881
- 4. Just RS, Scheible MK, Fast SA, Sturk-Andreaggi K, Röck AW, Bush JM, Higginbotham JL, Peck MA, Ring JD, Huber GE, Xavier C, Strobl C, Lyons EA, Diegoli TM, Bodner M, Fendt L, Kralj P, Nagl S, Niederwieser D, Zimmermann B, Parson W, Irwin JA (2015) Full mtGenome reference data: development and characterization of 588 forensic-quality haplotypes representing three U.S. populations. Forensic Sci Int Genet 14:141–155
- Børsting C, Morling N (2015) Next generation sequencing and its applications in forensic genetics. Forensic Sci Inte Genet 18:78–89
- Clarke AC, Prost S, JAL S, WTJ W, Kaplan ME, Matisoo-Smith EA, The Genographic Consortium (2014) From cheek swabs to consensus sequences: an A to Z protocol for high-throughput DNA sequencing of complete human mitochondrial genomes. BMC Genomics 15:68
- ENFSI DNA Working Group (2010) Recommended minimum criteria for the validation of various aspects of the DNA profiling process, Issue no 001
- Scientific Working Group on DNA Analysis Methods SWGDAM (2016) Validation guidelines for DNA analysis methods
- Parson W, Gusmão L, Hares DR, Irwin JA, Mayr WR, Morling N, Pokorak E, Prinz M, Salas A, Schneider PM, Parsons TJ (2014) DNA Commission of the International Society for Forensic Genetics: revised and extended guidelines for mitochondrial DNA typing. Forensic Sci Int Genet 13:134–142
- Scientific Working Group on DNA Analysis Methods SWGDAM (2019) Interpretation guidelines for mitochondrial DNA analysis by forensic DNA testing laboratories
- Parson W, Strobl C, Huber G, Zimmermann B, Gomes SM, Souto L, Fendt L, Delport R, Langit R, Wootton S, Lagacé R, Irwin J (2013) Evaluation of next generation mtGenome sequencing using the Ion Torrent Personal Genome Machine (PGM). Forensic Sci Int Genet 7:543–549
- Strobl C, Eduardoff M, Bus MM, Allen M, Parson W (2018) Evaluation of the precision ID whole MtDNA genome panel for forensic analyses. Forensic Sci Int Genet 35:21–25
- Woerner AE, Ambers A, Wendt FR, King JL, Moura-Neto RS, Silva R, Budowle B (2018) Evaluation of the precision ID mtDNA whole genome panel on two massively parallel sequencing systems. Forensic Sci Int Genet 36:213–224
- 14. King JL, LaRue BL, Novroski NM, Stoljarova M, Seo SB, Zeng X, Warshauer DH, Davis CP, Parson W, Sajantila A, Budowle B (2014) High-quality and high-throughput massively parallel sequencing of the human mitochondrial genome using the Illumina MiSeq. Forensic Sci Int Genetics 12:128–135
- McElhoe JA, Holland MM, Makova KD, Su MS, Paul IM, Baker CH, Faith SA, Young B (2014) Development and assessment of an optimized next-generation DNA sequencing approach for the mitogenome using the Illumina MiSeq. Forensic Sci Int Genet 13: 20–29
- Peck MA, Brandhagen MD, Marshall C, Diegoli TM, Irwin JA, Sturk-Andreaggi K (2016) Concordance and reproducibility of a next generation mtGenome sequencing method for high-quality

samples using the Illumina MiSeq. Forensic Sci Int: Genetics 24: 103–111

- Peck MA, Sturk-Andreaggi K, Thomas JT, Oliver RS, Barritt-Ross S, Marshall C (2018) Developmental validation of a Nextera XT mitogenome Illumina MiSeq sequencing method for high-quality samples. Forensic Sci Int Genet 34:25–36
- 18. Qiagen (2014) EZ1® DNA Investigator® Handbook
- 19. Qiagen (2014) QIAamp® DNA Micro Handbook
- National Institute of Standards and Technology NIST (2018) Certificate of Analysis: Standard Reference Material® 2392 Mitochondrial DNA Sequencing (Human).
- National Institute of Standards and Technology NIST (2018) Certificate of Analysis: Standard Reference Material® 2392-I Mitochondrial DNA Sequencing (Human HL-60 DNA)
- 22. Illumina (2016) Protocol: Human mtDNA Genome for the Illumina Sequencing Platform, Document #15037958 v01
- TaKaRa Bio Inc. (2015) PrimerSTAR® GXL DNA Polymerase Product Manual, Cat. #R050A, v201509Da
- 24. Illumina (2016) MiSeq® System Denature and Dilute Libraries Guide, Document #15039740 v01
- Illumina (2019) NextSeq® System Denature and Dilute Libraries Guide, Document #15048776 v10
- Barbarić L, Lipovac K, Sukser V, Rožić S, Korolija M, Zimmermann B, Parson W (2020) Maternal perspective of Croatian genetic diversity. Forensic Sci Int Genet 44:102190
- 27. Illumina (2016) mtDNA Variant Processor v1.0 BaseSpace App Guide, Document #100000007931 v00
- Anderson S, Bankier AT, Barrell BG, de Bruijn MH, Coulson AR, Drouin J, Eperon IC, Nierlich DP, Roe BA, Sanger F, Schreier PH, Smith AJ, Staden R, Young IG (1981) Sequence and organization of the human mitochondrial genome. Nature 290(5806):457–465
- Andrews RM, Kubacka I, Chinnery PF, Lightowlers RN, Turnbull DM, Howell N (1999) Reanalysis and revision of the Cambridge reference sequence for human mitochondrial DNA. Nat Genet 23(2):147
- Gilder JR, Doom TE, Inman K, Krane DE (2007) Run-specific limits of detection and quantitation for STR-based DNA testing. J Forensic Sci 52(1):97–101
- Riman S, Kiesler KM, Borsuk LA, Vallone PM (2017) Characterization of NIST human mitochondrial DNA SRM-2392

and SRM-2392-I standard reference materials by next generation sequencing. Forensic Sci Int Genet 29:181–192

- Robinson JT, Thorvaldsdóttir H, Winckler W, Guttman M, Lander ES, Getz G, Mesirov JP (2011) Integrative genomics viewer. Nat Biotechnol 29(1):24–26
- Robinson JT, Thorvaldsdóttir H, Wenger AM, Zehir A, Mesirov JP (2017) Variant review with the integrative genomics viewer (IGV). Cancer Res 77(21):31–34
- 34. Illumina (2015) MiSeq® System Specification Sheet
- 35. Illumina (2019) Cluster Optimization: Overview Guide, Document #1000000071511 v00
- Hussing C, Kampmann ML, Smidt Mogensen H, Børsting C, Morling N (2018) Quantification of massively parallel sequencing libraries – a comparative study of eight methods. Sci Rep 8:1110
- Loman NJ, Misra RV, Dallman TJ, Constantinidou C, Gharbia SE, Wain J, Pallen MJ (2012) Performance comparison of benchtop high-throughput sequencing platforms. Nat Biotechnol 30(5):434– 439
- Quail MA, Smith M, Coupland P, Otto TD, Harris SR, Connor TR, Bertoni A, Swerdlow HP, Gu Y (2012) A tale of three next generation sequencing platforms: comparison of Ion Torrent, Pacific Biosciences and Illumina MiSeq sequencers. BMC Genomics 13: 341
- Schirmer M, Ijaz UZ, D'Amore R, Hall N, Sloan WT, Quince C (2015) Insight into biases and sequencing errors for amplicon sequencing with the Illumina MiSeq platform. Nucleic Acids Res 43(6):e37
- 40. Ring JD, Sturk-Andreaggi K, Peck MA, Marshall C (2017) A performance evaluation of Nextera XT and KAPA HyperPlus for rapid Illumina library preparation of long-range mitogenome amplicons. Forensic Sci Int Genet 29:174–180
- Brandhagen MD, Just RS, Irwin JA (2020) Validation of NGS for mitochondrial DNA casework at the FBI laboratory. Forensic Sci Int Genet 44:102151

Publisher's note Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.