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# PSR: polymorphic SSR retrieval

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# **Abstract**

**Background:** With the advent of high-throughput sequencing technologies large-scale identification of microsatellites became affordable and was especially directed to non-model species. By contrast, few efforts have been published toward the automatic identification of polymorphic microsatellites by exploiting sequence redundancy. Few tools for genotyping microsatellite repeats have been implemented so far that are able to manage huge amount of sequence data and handle the SAM/BAM file format. Most of them have been developed for and tested on human or model organisms with high quality reference genomes.

**Results:** In this note we describe *polymorphic SSR retrieval (PSR)*, a read counter and simple sequence repeat (SSR) length polymorphism detection tool. It is written in Perl and was developed to identify length polymorphisms in perfect microsatellites exploiting next generation sequencing (NGS) data. *PSR* has been developed bearing in mind plant non-model species for which *de novo* transcriptome assembly is generally the first sequence resource available to be used for SSR-mining. *PSR* is divided into two modules: the read-counting module (*PSR\_read\_retrieval*) identifies all the reads that cover the full-length of perfect microsatellites; the comparative module (*PSR\_poly\_finder*) detects both heterozygous and homozygous alleles at each microsatellite *locus* across all genotypes under investigation. Two threshold values to call a length polymorphism and reduce the number of false positives can be defined by the user: the minimum number of reads overlapping the repetitive stretch and the minimum read depth. The first parameter determines if the microsatellite-containing sequence must be processed or not, while the second one is decisive for the identification of minor alleles. *PSR* was tested on two different case studies. The first study aims at the identification of polymorphic SSRs in a set of *de novo* assembled transcripts defined by RNA-sequencing of two different plant genotypes. The second research activity aims to investigate sequence variations within a collection of newly sequenced chloroplast genomes. In both the cases PSR results are in agreement with those obtained by capillary gel separation.

**Conclusion:** *PSR* has been specifically developed from the need to automate the gene-based and genome-wide identification of polymorphic microsatellites from NGS data. It overcomes the limits related to the existing and time-consuming efforts based on tools developed in the pre-NGS era.

**Keywords:** Simple sequence repeats, Length polymorphism, Polymorphic microsatellites, NGS, SAM/BAM format

# **Background**

Simple sequence repeat (SSR), also known as short tandem repeats (STRs), variable number tandem repeats (VNTRs) or microsatellite markers, are repetitive stretches of 1–6 nucleotide units randomly spread in eukaryotic genomes. Microsatellites typically vary in length between 5 and 40 repeat units and can be classified

as perfect or imperfect: in the first case, the repeat unit is arranged in head to tail manner without any interruption; in the second case the repeat unit includes insertions, deletions or substitutions of bases. In addition, SSRs can be of two types: simple or compound (i.e. two adjacent distinct SSRs are separated by a certain number of nucleotides). A further categorization allows SSRs to be recognized as genomic SSRs (gSSRs) and expressed sequence tag (EST)-SSRs; the latter result from transcribed regions of a genome, are associated with functionally annotated genes and are supposed to be more conserved among related species/genera. SSRs are very polymorphic due

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to the high mutation rate affecting the number of repeat units and have several advantages over other molecular markers: they allow the identification of multiple alleles at single *locus*, are equally distributed all over the genome, show a co-dominant inheritance pattern and could be mapped in different populations becoming the "backbone" of high resolution linkage maps. The development of EST-SSR markers was historically based on collection of sequence data of complementary DNAs (cDNAs) generated using the Sanger sequencing technology. Some useful tools have been made available for the detection and localization of both perfect and imperfect microsatellites along genome and mRNA sequences [1–6] (http://www.bioinformatics.org/ftp/pub/msatfinder/; http://pgrc.ipk-gatersleben.de/misa/).

Generally, these tools are equipped with additional programs able to design primers for microsatellite flanking regions. One of the most common approaches to the detection of length polymorphisms relies on the random selection of SSRs and on the analysis of PCR products on high-resolution gels [7]. Since the application of this method is labour-intensive and time-consuming, it is more advisable to adopt an in silico approach to identify polymorphic SSRs and subsequently validate them through capillary gel separation. Relatively, little work has been done till now toward the automatic identification of polymorphic SSRs by exploiting EST sequence redundancy [4]. With the advent of next-generation sequencing (NGS) methods, the large-scale identification of microsatellites became rapid and cost-effective [8] and was especially directed to non-model species with un-sequenced genomes [9–14]. But, high-throughput NGS technologies posed new challenges to existing bioinformatics applications that need to face a huge amount of sequence data and to handle new file formats. Until a few years ago, no adequate tools have been developed for the identification of polymorphic SSRs from NGS data with the exception of custom scripts still based on the detection of pad characters in the output of the final CAP3 assembly [9] or of the use of graphical viewer for NGS assemblies as starting point to manually explore and estimate microsatellite variability at each SSR *locus* [12, 15, 16]. It is clear, therefore, that it is no longer possible to work with humanreadable text files including multiple sequence alignments of reads to the reference; rather it is essential to consider compressed text files in SAM/BAM format [17] to analyse sequence alignments and extract useful patterns. As far as we know, lobSTR [18], RepeatSeq [19] and the recently developed STR-FM [20] are the mainly publically available tools for profiling microsatellites from SAM/BAM data. All of them were conceived to identify gSSR alleles at each *locus* bearing in mind the short size of NGS reads and the difficulties this type of sequence data arises in the correct identification of polymorphic SSRs. An additional effort has been freshly published [15], but the method described is not publically available. However, these tools, even though capable of genotyping also EST-SSRs, were developed for and tested on human or model organisms for which the genome is available and represents a good reference for SSR-mining. Indeed, it seems that their adaptability is not likely to allow the easy use of custom reference transcriptomes and/or automatic comparative analysis at microsatellite loci. In this paper, we present polymorphic SSR retrieval (PSR) a Perl package developed to identify polymorphic SSRs from NGS data and provide quantitative information to each call. Unlike the aforementioned tools, PSR has been developed in a context where de novo transcriptome assembly is generally the first sequence resource for non-model plant species to be used for SSR-mining and downstream investigations.

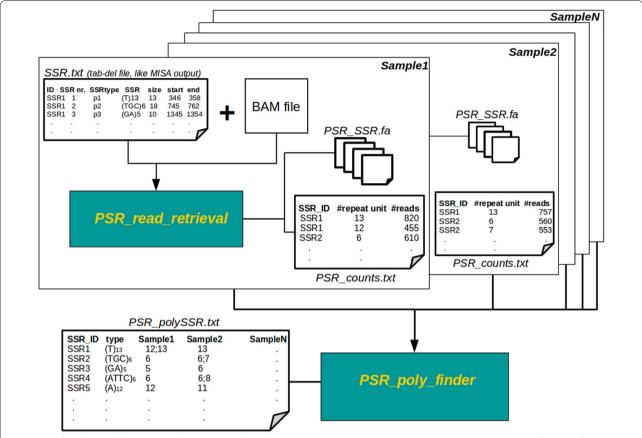
# **PSR** workflow

*PSR* is a Perl package, conceived as a modular and flexible tool, developed to identify polymorphic SSRs from NGS data. The *PSR* package and the user manual are available at the following URL: <a href="http://sourceforge.net/projects/polyssr/">http://sourceforge.net/projects/polyssr/</a>.

The user guide provides general information on software dependencies and installation procedure as well as detailed instructions for running the application. PSR workflow is shown in Fig. 1 and the text that follows describes the key points. Psr\_read\_retrieval aims at the identification of all the reads that align to the reference sequences covering the full-length of perfect microsatellites. We decided to focus on perfect microsatellite only, because the total length of the reads obtained by the Illumina sequencing instruments rarely exceeds the 100 high quality nucleotides. These constraints appear to limit polymorphism discovery only to short polymorphic SSRs, but as it is evident from data in Table 1, the maximum length of the microsatellite is always slightly lower than the one of the sequenced read. In addition, the number of reads covering each microsatellite does not reflect the total number of reads at each locus. Indeed, in case SSRs are located at the ends of the read, the microsatellite-containing sequences are discarded since they can strongly affect the call of polymorphic sites (Fig. 2).

In addition, *PSR* filters out all the reads that match twice or more on the reference sequence as well as non-overlapping paired-end reads that aligned on the same microsatellite. This filtering step allows the user to exclude misalignments that can greatly influence downstream analysis.

Sometimes it could happen that the most frequent microsatellite identified by *PSR* presents a number of repeat units that differs from the one detected on the



**Fig. 1** Workflow design of the polymorphic SSR retrieval tool that includes two modules. *PSR\_read\_retrieval* aims at the identification of all the reads that cover the full-length of perfect microsatellites. N indicates the number of iteration that must correspond to the number of genotypes under investigation. *PSR\_poly\_finder* detects length polymorphism in microsatellites

Table 1 The average read length obtained from Illumina instruments is typically less than 150 nucleotides

Repeat unit (nts)	Max number of repeat unit	SSR length (nts)					
Read length = 100 nucleotides (nts)							
2	48	96					
3	31	94					
4	23	92					
5	18	90					
6	14	88					

This length allows to detect di-nucleotide repetition up to 96 nucleotides and 14 repetitions of esa-nucleotide pattern

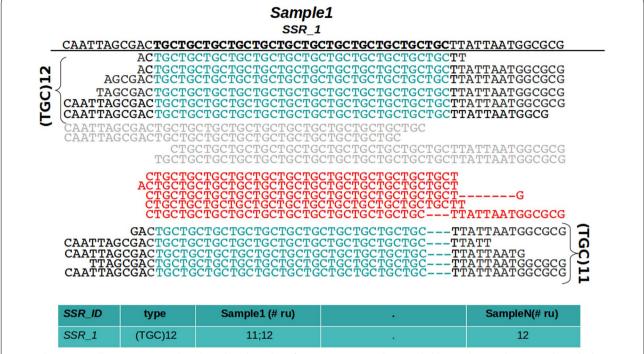
microsatellite-containing reference sequence. This is not surprising since the reference sequence was assembled considering all the reads; by contrast *PSR* takes into account only those reads that entirely overlap the repeating pattern.

*Psr\_read\_retrieval* takes two files as input: a 7-column tab-delimited file and a BAM file. The first one is

a MISA-like output file (http://pgrc.ipk-gatersleben.de/misa/misa.html) that includes the type and the size of identified microsatellites and their localization within the reference sequences. The second input file stores read alignments against reference sequences for each genotype under investigation. The latter can be genome sequences or *de novo* assembled transcripts depending on the type of microsatellites under study.

Based on our experience and on results by Highnam et al. [19], we suggest bowtie2 [21] as reference sequence alignment mapping tool. However, we are well-aware that read mapping tools are constantly being developed; for that reason we believe more convenient to stay independent from alignment algorithm.

Psr\_read\_retrieval should be run as many times as the number of genotypes under investigation (N in Fig. 1). It generates two different outputs: a collection of FASTA files including all the reads that successfully aligned along each SSR and different 3-column tab-delimited files, one for each genotype under investigation, where for each microsatellite the number of repeat units and



**Fig. 2** The microsatellite region (36 nucleotides in length) in the reference sequence is shown in *bold. PSR* selects and counts only reads aligning the entire SSR region, in addition to one or more nucleotides to the *left* and *right side* different from bases in the repetitive units. Reads in *grey* are discarded since they partially cover the microsatellite. Reads in *red* are filtered out due to the uncertainty at the sequence borders. For each sample the program returns the number of repetitive units presenting coverage greater than the selected threshold (—p) in a tab delimited file

the number of supporting reads are reported. In case of genomic DNA, results from this step of the analysis allowed ambiguities associated with repetitive stretches to be fixed as well as sequence assembly quality to be assessed.

Then, all the table-wise text files are fed into the *psr\_poly\_finder* module that, exploiting MySQL DBMS, creates a query table joining all the info recorded in text files in order to detect both heterozygous and homozygous alleles at each microsatellite *locus*. Users can define two different parameters, namely the minimum number of supporting reads (-t) and the minimum read depth (-p), for length polymorphism to be called. By default the values of these parameters are fixed to 10 and 30, respectively.

To give a clear example, a microsatellite is analysed in case at least ten reads overlap the repetitive stretch and it is called as heterozygous if 30 % or more of the reads supports the identification of the minor allele. In case of polyploid genotypes, the user can fine-tune the value of the "p" parameter in order to support the detection of all the putative alleles at each *locus*.

Filtering by a minimum number of reads supports variant calls and by minimum read depth within the microsatellite interval helps to reduce false positive calls.

Psr\_poly\_finder produces as final output a table-wise text file where the first two columns report the identifiers of microsatellite-containing sequences and the type of the identified microsatellite; the remaining columns include one or more (in case of heterozygosity) SSR allele sizes per genotype.

In order to evaluate *PSR* reliability and effectiveness, the package has been tested in the identification of polymorphic SSRs in (1) a collection of *de novo* assembled transcripts defined by RNA-sequencing of two different plant genotypes; (2) a set of newly sequenced chloroplast genomes.

In the first case, ten randomly selected polymorphic microsatellites, as they were identified by *PSR*, were assessed by capillary electrophoresis and eight of them showed the expected length polymorphism (Table 2). Only one SSR *locus* is heterozygous in case of genotype 1 and results from in silico and wet-lab experiments are in agreement. As for genotype 2, seven out of ten microsatellite *loci* resulted heterozygous even if *PSR* identifies both the alleles (major and minor) only in five cases.

In the latter case *PSR* combined with capillary gel separation was used to assess whether in silico identified length polymorphisms represented sequencing/assembling errors or, alternatively, real genotype-specific

Table 2 Evaluation through capillary gel separation of SSR length polymorphism from ten randomly selected transcripts

Seq ID	SSR	Start	Stop	SSR length	SSR type	Capillary electrophoresis		PSR	
						Genotype 1	Genotype 2	Genotype 1	Genotype 2
TR11073	(GAT)6	389	406	18	р3	222	222 + 219	6	6; 5
TR11727	(TCA)5	1026	1040	15	р3	182	179	6	5
TR12365	(GAT)6	269	286	18	р3	243 + 246	243 + 246	8; 6	8; 6
TR19012	(CAC)7	208	228	21	р3	239	233 + 239	7	7; 5
TR6251	(CAA)9	204	230	27	р3	294	294 + 297	9	10; 9
TR1824	(CAT)5	90	104	15	р3	277	274	6	5
TR12469	(TC)7	1300	1313	14	p2	234	232	7	6
TR7469	(ATC)7	20	40	21	р3	237	237 + 240	7	7; 8
TR2455	(TTG)6	57	74	18	р3	239	239 + 242	6	6
TR142	(TCT)9	1874	1900	27	р3	159	156 + 159	9	9

Columns 7 and 8 report amplicon size detected into two genotypes. Columns 9 and 10 list the number of repeated units as identified by PSR

sequence variations. In seven cases out of eight, the size of the microsatellite supported by the highest number of reads was confirmed by capillary electrophoresis (Table 3). The use of *PSR* proved to be particularly useful to fix sequencing errors in monomeric microsatellites.

It is very likely that differences from in silico and wet-lab experiments depend on the amount of reads *PSR* uses to determine the number of repeated unit and microsatellite length. Indeed, for each SSR region *PSR* discards all the reads (on average the 60 % of the total) that do not completely span the microsatellite. Furthermore, in case of EST-SSRs, the quality of the reference transcriptome may negatively affect read alignment as a first step and, subsequently, read count. Finally, differences can be the result of PCR slippage products of repetitive stretches or can be ascribed to errors generated by PCR-based NGS technologies, especially in case of monomeric microsatellites [20].

#### **Conclusions**

The ability of next generation sequencing technologies to produce large amounts of sequence data accelerated microsatellite identification and facilitated the discovery of polymorphic SSRs. In addition, NGS resulted in the development of new specialised tools and file formats that support the management of huge amount of sequence data. However, as far as we know, few automatic and efficient systems have been developed so far for detecting both gene-based and genome-wide polymorphic microsatellites from NGS data. In this note we proposed PSR, a Perl Package developed for the detection of length polymorphisms and for the automatic comparative analysis at microsatellite *loci.* Indeed, the availability of polymorphic SSR markers is an essential requirement for plant community since it serve as basic resources for QTL mapping, molecular breeding, genetic diversity analysis and SSR fingerprinting for food traceability. All these and further applications, motivated us to develop and distribute *PSR* within plant community.

Table 3 Evaluation through capillary electrophoresis of eight monomeric SSR loci across nine cpDNA genotypes

Sample	SSR1 (T)n	SSR2 (T)n	SSR3 (T)n	SSR4 (A)n	SSR5 (T)n	SSR6 (A)n	SSR7 (A)n	SSR8 (A)n
G1	11	13	16	12	16	9	16	18
G2	13	13	16	12	13	9	13	16
G3	12	13	16	13	17	10	16	15
G4	12	13	16	13	17	10	16	15
G5	12	13	16	13	17	10	16	15
G6	12	13	16	13	17	10	16	15
G7	12	13	16	13	17	10	16	15
G8	12	13	16	13	17	10	16	15
G9	12	13	16	13	17	10	15	15

Numbers represent the length of SSR stretches. Italics cells indicate microsatellites that have been confirmed also by Sanger sequencing. Bold italics cells represent SSRs with different lengths compared to those determined by PSR

# **Availability and requirements**

Project name: PSR—polymorphic SSR retrieval Project home page: http://sourceforge.net/projects/ polyssr/

Operating system(s): tested on Linux and Mac OS X

Programming language: Perl

Other requirements: BioPerl, MySQL

License: GNU GPL 3

Restrictions to use by non-academics: None.

#### **Abbreviations**

PSR: polymorphic SSR retrieval; SSR: simple sequence repeat; STR: short tandem repeat; VNTR: variable number tandem repeats; gSSRs: genomic SSRs; EST-SSRs: expressed sequence tag-SSR; NGS: next generation sequencing; QTL: quantitative trait *loci*; DBMS: database management system.

#### Authors' contributions

CC and NDA designed the core of the Perl package. CC developed the package and contributed to the writing of the manuscript. NDA supervised PSR implementation and wrote the manuscript. All authors read and approved the final manuscript.

## Acknowledgements

We sincerely thank Nunzia Scotti (CNR-IBBR), for sharing plastid genome-scale data and Andrea Allavena (CREA-FSO) for sharing RNA-sequencing data. We also thank Teodoro Cardi (CREA-ORT) for critical reading of the manuscript. This work was carried out in the frame of the "GenHORT—adding value to elite Campania horticultural crops by advanced genomic technologies" Project (PON02\_00395\_3215002) and is supported by the PON R&C 2007–2013 grant funded by the Italian Ministry of Education, University and Research in cooperation with the European Funds for the Regional Development (FESR).

## Compliance with ethical guidelines

#### **Competing interests**

The authors declare that they have no competing interests.

Received: 6 March 2015 Accepted: 21 September 2015 Published online: 01 October 2015

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