RESEARCH ARTICLE



Microsatellites to enhance characterization, conservation and breeding value of *Capsicum* germplasm

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Abstract Capsicum species are of great economic, medicinal, culinary and cultural value worldwide, however their genetic potential is still underutilized. In this study, a set of 21 microsatellite markers showed utility for distinguishing a diverse group of 42 Capsicum genotypes representing 11 species. The level of polymorphism among these genotypes was evaluated by calculating the polymorphic information content (PIC) and discriminating power (D) for each locus. A total of 208 alleles were detected at the 21 loci evaluated. Markers detecting the highest number of alleles also exhibited the highest number of rare alleles. PIC values and D estimations showed a significant positive correlation (r = 0.98)P < 0.0000). PIC values ranged between 0.500 (marker NG20) and 0.899 (marker NG17), giving an average of 0.775. Genetic relationships among genotypes were evaluated using UPGMA clustering, Principal Coordinate Analysis (PCoA) and Bayesian model-based clustering, allowing genotypes to be discriminated into groups according to their degree of genetic similarity corresponding to distinct taxa. Based on these analyses, the 21 microsatellite included in this study now define a standard tool for efficient assessment of molecular variability of *Capsicum* germplasm collections conserved worldwide.

Keywords Cross-amplification · Genetic erosion · Germplasm banks worldwide · SSR markers usefulness

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Introduction

The fruits of the *Capsicum* (pepper) genus are valued worldwide as both a vegetable and spice. Pepper ranks fourth on the list of the most important vegetables in the global food economy, after tomatoes, cucurbits (pumpkins, squashes, cucumbers, and gherkins) and alliums (onions, shallots, and garlic) (Schreinemachers et al. 2018). In the United States, the 2017 fresh market value of bell and chili pepper crops was over \$750 million dollars (National Agricultural Statistics Service 2018). The *Capsicum* genus is native to South and Central America (Bosland 1996; Perry et al. 2007;



Walsh and Hoot 2001) and comprises approximately 35 species, of which five are domesticated and widely consumed worldwide as both vegetable foods and spices (Carrizo García et al. 2016; Moscone et al. 2007). This number, however, is only an estimate as new species are discovered and named on a regular basis (Barboza et al. 2019; Samuels 2015).

The five domesticated species are Capsicum annuum L., C. baccatum L., C. chinense Jacq., C. frutescens L., and C. pubescens Ruiz and Pav. Two centers of domestication of the genus have been identified: one in Mexico for C. annuum and the other one in South America for the other species (Eshbaugh 1993). One of pepper's most notable characteristics is its pungency, attributed to a unique family of composite alkaloids produced in the fruit known as capsaicinoids (Bosland 1999). Most peppers grown commercially worldwide belong to the species C. annuum [including sweet varieties (e.g. bell peppers and paprika) and hot varieties like cayenne, jalapeños, and the chiltepin] with a few additional representatives from C. chinense (e.g. Habanero) and C. frutescens (e.g. Tabasco). In turn, C. baccatum and C. pubescens include the South American ají and rocoto peppers, respectively. Pepper fruits are an excellent source of micronutrient antioxidants (vitamins C and E as well as carotenoids), which confer to Capsicum a trending nutraceutical value due to their role in preventing or reducing chronic and age-related diseases. (Materska and Perucka 2005; Palevitch and Craker 1996; Soare et al. 2017).

Despite their broad presence in the world diet and global economic importance, the taxonomic classification of Capsicum within and between species is still confusing (Barboza 2011; Eshbaugh 1993). The terms pepper, chili, chile, and Capsicum are used interchangeably for different types and species within the genus (Andrews 1996). Additionally, there is no clear delimitation within two complexes of Capsicum species: (1) the C. annuum complex, formed by C. annuum, C. frutescens, and C. chinense, and (2) the C. eximium complex, formed by C. eximium Hunz. and C. cardenasii Heiser and P.G.Sm. (Moscone et al. 2007; Walsh and Hoot 2001). Ultimately, wild Capsicum germplasm represents an important source of genetic diversity but is underutilized and worthy of further research (van Zonneveld et al. 2015).

Currently, there exist several large germplasm ex situ collections of both wild and domesticated

Capsicum species, particularly those maintained by the World Vegetable Center (AVRDC, Taiwan) with 8170 accessions, the United States Department of Agriculture (USDA) with 6067 entries, and the National Bureau of Plant Genetic Resources (NBPGR, India) with 2774 accessions (Chhapekar et al. 2018). Universities and other institutions from 21 countries in Latin America and the Caribbean also have genebanks that conserve smaller collections of Capsicum germplasm (Knudsen 2000). Nevertheless, only a few of these collections have been even partially characterized for morphological characters. Certain Capsicum species are not represented in the collections at all and perhaps will never be included because agricultural development and expanding human populations are reducing their natural habitats (Bosland and Gonzalez 2000).

Morphological characteristics have been traditionally used to classify species; however, they alone are not always enough to adequately categorize a specific entry to the species level in an ex situ collection. This type of characterization requires extensive resources, such as appropriate facilities and time (plants must be grown to the flowering or fruiting stage), as well as special expertise to determine characteristics such as pollen shape and flower morphology (IBPGR 1983; Singh et al. 2004). Alternatively, molecular assessment of germplasm collections may guide decisions about conservation activities, such as collecting, managing, identifying genes, and adding value to genetic resources (de Vicente et al. 2006).

Despite current advances in genome sequencing, microsatellite markers or simple sequence repeats (SSR) that have been converted into a DNA marker are still successfully used worldwide for routine genotyping in many crops, and for protection of ancient autochthonous olive trees in Italy (Rotondi et al. 2018), evaluation of genetic redundancy in ex situ apple collections in Norway (Gasi et al. 2016), evaluation of genetic structure in subspecies of black cherry from North America (Guzmán et al. 2018), estimation of natural outcrossing rate and genetic diversity in Lima bean landraces from Brazil (Penha et al. 2017), and in the analysis of the genetic diversity of rice cultivars in India (Dhama et al. 2018). This molecular technique is also ideal for designing conservation strategies for genetic resources, and for forming and validating core collections (Porta et al. 2018; de Vicente et al. 2006).



Capsicum germplasm collections are of actual and potential value in the breeding of the genus, and microsatellites are useful in the exploitation of these valuable sources of genetic diversity. Thus, the aims of this study were to (1) assess the utility of a collection of 54 microsatellite markers (Lee et al. 2004; Nagy et al. 2007) for distinguishing a diverse panel of 42 Capsicum genotypes, and (2) provide a rapid, accurate, and cost-effective set of SSR markers for assessment of genetic diversity in germplasm collections.

Materials and methods

Plant material and DNA extraction

This study included 42 genotypes representing 11 species of the *Capsicum* genus (Table 1). Germplasm was obtained from commercially available varieties or obtained from the USDA collection based on the USDA Genetic Resources Information Network (GRIN). At least three seeds per accession were planted under standard greenhouse conditions (16 h light and eight hours darkness with weekly fertilization). Young meristematic leaves were collected, and DNA was extracted according to Doyle and Doyle (1987). The leaf tissue of all individuals belonging to the same genotype was pooled to extract DNA. The quality of the extracted DNA was assessed by agarose gel electrophoresis (1% agarose) using a TBE 0.5X (Tris-borate-EDTA) buffer and staining with ethidium bromide. The gel was exposed to ultraviolet light and photographed. DNA was diluted to a final concentration of 2.0 ng/ μ L and stored at -20 °C until use.

Microsatellite amplification

Fifty-four microsatellite markers were selected for preliminary evaluation. Thirty-three SSR were eliminated from the study because they exhibited poor band resolution. The remaining 21 microsatellite markers were then used to evaluate molecular diversity of the 42 genotypes included in this study (Table 2).

Fifteen SSRs have an NG designation, as they were initially reported by Nagy et al. (2007). Amplification was performed in a total volume of 15.0 μ L containing 5.0 ng DNA, 0.5 μ M of each primer, 0.125 mM of

each dNTP, 1.5 μ L 10X buffer (0.1 M Tris pH 8.3, 0.5 M KCl, 7.5 mM MgCl₂, 0.1% gelatin), and 1 unit of Taq polymerase. The remaining six SSRs have a designation BD, as they were first reported by Lee et al. (2004), and their amplification was performed in a final volume of 10.0 μ L containing 6.0 ng DNA, 0.3 μ M of each primer, 0.1 mM of each dNTP, 1.0 μ L 10X buffer (0.1 M Tris pH 8.3, 0.5 M KCl, 7.5 mM MgCl₂, 0.1% gelatin), and 1 unit of Taq polymerase. Negative controls were run with every PCR to ensure no DNA contamination.

All amplifications were performed using a PTC-225 programmable thermocycler (MJ Research Inc., Waltham, MA, USA) under the conditions indicated in Table 2. Several amplifications were performed doing "Touchdown" PCR to reduce the amplification of non-specific products (Don et al. 1991). Touchdown amplification reactions began at a high hybridization temperature that gradually decreased by 1 °C per cycle until the respective hybridization temperatures of primers were reached.

The electrophoresis of amplified products was performed on 4% denaturing polyacrilamide gels, with 7 M urea and TBE 0.5X buffer. Six microliters of buffer [95% (v/v) formamide, 20 mM EDTA, pH 8.0, 0.05% (w/v) bromophenol blue and xylene cyanol FF] were added to each PCR product. This mixture was heated at 95 °C for 5 min and immediately cooled down on ice, then 5.0 μ L of each sample were loaded into the wells of the gel.

The samples were run (2000 V, 75 W, 50 mA) and stained immediately afterwards with silver nitrate. Table 2 shows the electrophoresis run time for each marker. After staining with silver nitrate, discreet groups of two to five bands were observed (what is commonly called 'shadow bands' or 'stuttering') in most of the markers. 'Stuttering' refers to PCR products differing in the size of the principal band (allele), which makes each allele (band) observed in the gel appear to be more than one band (Litt et al. 1993).

Twenty-seven unique samples were run on each gel as a check to evaluate the reproducibility of patterns and compare bands between gels. Fifteen genotypes were also chosen at random and subsequently amplified separately and run several times on different gels.

The size of the most intensely amplified band of each microsatellite, measured as the number of nucleotides, was determined based on its migration



Table 1 List of the 42 Capsicum genotypes used in this study

	Species	Code	Cluster		Genotype/accession	Pungency	
			UPGMA	Structure			
1	C. annuum	1ann	A	C1	Maor ^a	Sweet	
2	C. annuum	2ann	A	C1	Jupiter ^a	Sweet	
3	C. annuum	3ann	A	C1	Jalapeño 3575 ^a	Sweet	
4	C. annuum	4ann	A	C1	Jalapeño 1493 ^a	Spicy	
5	C. annuum	5ann	A	C1	Jalapeño M ^a	Spicy	
6	C. annuum	6ann	A	C1	RNaky ^a	Spicy	
7	C. annuum	7ann	A	C1	PapriQueen ^a		
8	C. annuum	8ann		C6	Thai hot ^a	Spicy	
9	C. annuum	9ann	A	C1	Chung yan ^a	Spicy	
10	chinense x annuum	10chixann		C6	Bahamian Superhot ^a	Spicy	
11	C. chinense	11chi	В	C2	Habanero 11 ^a	Spicy	
12	C. chinense	12chi	В	C2	Peach Savina ^a		
13	C. chinense	13chi	В	C2	NMCA30036	Sweet	
14	C. chinense	14chi	В	C2	PI159234		
15	C. frutescens	15fru	C	C3	Tabasco ^a	Spicy	
16	C. frutescens	16fru		C7	2814-6-2	Spicy	
17	C. annuum	17ann	A	C1	Cayenne ^a	Spicy	
18	C. chacoense	18cha	E	C9	PI439414		
19	C. chacoense	19cha	E	C9	PI555612		
20	C. baccatum	20bac			var. praetermissum		
21	C. pubescens	21pub	E	C5	80062	Spicy (decline)	
22	C. pubescens	22pub	E	C5	PI158277	Sweet (spicy very low)	
23	C. minutiflorum	23 min	E	C8	Wild	Spicy	
24	C. cardenasii	24car	E	C7	Wild		
25	C. flexuosum	25flex	Е	C7	Wild	Sweet	
26	C. rhomboideum	26rho	F	C10	Wild	Sweet	
27	C. rhomboideum	27rho	F	C10	Wild	Sweet	
28	C. annuum var. abbreviatum	1abb	A	C6	PI93475011B-03-3		
29	C. baccatum	16bac	D	C4	NMCA 20016A		
30	C. luteum	28lut		C6	PI979750065A-03-1		
31	C. baccatum var. baccatum	37bacbac	D	C4	PI260567-3-26		
32	C. baccatum var. pendulum	43bacpen	D	C4	PI241679C-mm02-28		
33	C. baccatum var. pendulum	49bacpen	D	C4	PI24167919-02-47		
34	C. baccatum var. pendulum	51bacpen	D	C4	PI337522B-mm-02-34		
35	C. chinense	53chi	C	C3	PI281421B-03-9		
36	C. chinense	54chi	C	C3	PI281421B-03-9-A		
37	C. chinense	55chi	C	C3	PI281421B-03-9-B		
38	C. baccatum	56bac		C10	PI439378A-mm02-36		
39	C. frutescens	58fru	C	C3	PI441648A-0242		
40	C. frutescens	59fru	D	C4	PI441648A-0242-A		
41	C. baccatum	60bac	D	C4	PI439381A-mm02-39		



Table 1 continued

	Species	Code	Cluster		Genotype/accession	Pungency
			UPGMA	Structure		
42	C. frutescens	61fru	С	C3	PI446908A-03-17	_

The columns present the botanical species, the abbreviation used in data analyses, the resulting group from UPGMA and Bayesian model-based clustering analysis, name of genotype or accession number, and the pungency level

Table 2 Conditions of amplification and electrophoresis for the 21 SSR evaluated

SSR	Locus	Repeat	Primer	PCR ^a	ET	
			Forward	Reverse		
NG4	CA515275	(TC) ₄ (CT) ₁₃	ctctgccctcctcaaccc	aaaatatggtcggagatccg	55, 35	40
NG5	BM067271	(CTG) ₆	gtatgctgcaaccatcgttg	attggtttgggagacacagc	55, 35	75
NG6	CA515649	(CAT) ₆	tctccaatttccattcggag	taatcgcatttgcgaacttg	55, 35	60
NG7	CA525390	$(CT)_{14}(CA)_9GA(CA)_4GA(CA)_4$	ggaaactaaacacactttctctctc	actggacgccagtttgattc	55, 35	50
NG8	CA516044	$(CA)_{10}$	atettetteteattteteeette	tgctcagcattaacgacgtc	55, 35	45
NG9	BM061910	T_{16}	attgtgatagcaacccctgg	cacagatgagggcacaaatg	55, 35	70
NG10	CA516334	$(CAA)_6$	acccaccttcatcaacaacc	atttgtggcttttcgaaacg	55, 35	70
NG12	BM064867	(CAT) ₆	tctgggaattttggaactgc	tccagttttgatcatctccaac	55, 29 ^b	50
NG15	CA515055	$(CAT)_2(CAT)_6$	taatcgagcggtagattcgg	taagtggaggtgcccttctg	55, 35	90
NG17	GPMS93	$(TA)_{14imp}(GA)_{27imp}$	atccttggcgtattttgcac	ttcactttgcacacaggctt	55, 35	90
NG18	CA517699	$(CA)_{15}$	acgccaagaaaatcatctcc	ccattgctgaagaaaatggg	55, 29 ^b	40
NG20	BM061461	$(GA)_9$	ctcattaccacttcatacaaaacag	tgcagtaggtgttgctacgg	55, 35	60
NG33	CA516439	$(AG)_{11}$	gacagtctttcaagaactagagagag	tggagcaaacacagcagaac	55, 35	50
NG34	GPMS117	$(TA)_{25}(GA)_{14}$	gatgttaggtccgtgcttcg	aagccccatggaagttatcc	55, 29 ^b	45
NG35	GPMS187	$(CA)_6C_4A_{36}$	tttagaatcctcaccacggg	tcaatgcacaaactttaatttgc	55, 29 ^b	60
BD12	Hpms 1-139	$(CT)_2(AG)_{15}$	ccaacagtaggacccgaaaatcc	atgaaggctactgctgcgatcc	58, 29 ^c	80
BD17	Hpms 1-165	$(GA)_{13}$	ggctatttccgacaaaccctcag	ccattggtgttttcactgttgtg	54, 35	60
BD22	Hpms 1-214	$(GTTT)_2(TTG)_9$	tgcgagtaccgagttctttctag	ggcagtcctgggacaactcg	58, 35	40
BD33	Hpms 2-23	$(TTG)_7(GT)_9$	ccctcggctcaggataaatacc	ccccagactcccactttgtg	57, 35	40
BD41	HpmsCaSIG19	$(CT)_6(AT)_8(GTAT)_5$	catgaatttegtettgaaggteee	aagggtgtatcgtacgcagcctta	58, 29 ^c	60
BD70	CM0005	(CCA) ₈	catgaccaccatgaggata	gatagccacgagcatagtatt	54, 29 ^d	50

ET electrophoresis time in minutes

^aPCR conditions: annealing temperature (°C) and number of cycles, respectively

^bPCR touchdown: 63–1 °C/cycle ^cPCR touchdown: 65–1 °C/cycle ^dPCR touchdown: 62–1 °C/cycle

in relation to that of molecular weight markers (50 bp and 10 bp DNA ladders, Roche Diagnostics Corporation, Indianapolis, IN, USA). Due to their codominance, the markers were scored as homozygous or heterozygous.

SSR data analysis

Each intensely amplified SSR band was manually scored as present (1) or absent (0), and a binary data matrix was generated. Two value measures for each marker were calculated on an Excel spreadsheet. First, the

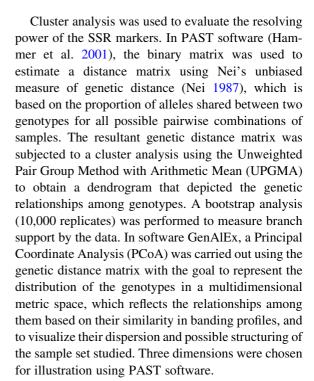


^aCommercial variety

Polymorphism Information Content (PIC) value was estimated according to Botstein et al. (1980): PIC = $1 - \left(\sum_{i=1}^{n} P_i^2\right) - \sum_{i=1}^{n-1} \sum_{j=i+1}^{n} \left(2P_i^2 P_j^2\right)$, where Piand Pj are the relative frequencies of the ith and jth alleles in a given marker, and n is the total number of alleles detected by the marker. PIC value estimates the discriminatory power of an SSR by taking into account the number of alleles exhibited and their relative frequencies. Second, the discriminating power (D) per marker was calculated according to Tessier et al. (1999): D = $1 - C_j = 1 - \sum_{i=1}^{I} P_i \frac{(NP_i - 1)}{N - 1}$, where *I* is the number of banding patterns (or genotypes) generated by the SSR, Pi is the frequency of the ith pattern of the given jth primer, N is the number of individuals evaluated and Cj is the confusion probability for the jth SSR, i.e. the probability that two randomly chosen individuals from the sample have identical banding patterns. In consequence, $D = 1 - C_i$ represents the probability that two randomly chosen individuals have different patterns and, therefore, are distinguishable from one another. Additionally, as N tends toward infinity, the limit of D, equals to D_L , provides an estimate of discriminating power of each primer as follows: $D_L = 1 - C = 1 - \sum_{i=1}^{I} P_i^2$ (Tessier et al. 1999).

Theoretically, in a set of N individuals, it is possible to draw (N(N-1)/2) different pairs. According to Tessier et al. (1999), the total number of non-differentiated pairs of genotypes for the jth SSR is given by $x_j = (N(N-1)/2)C_j$. Thus, for a given combination of K primers, $X_K = (N(N-1)/2)$ $\Pi_{j=1}^k C_j$ estimates the number of non-differentiated pairs of genotypes by the combination. The most efficient combination reduces the most the number of undifferentiated pairs, i.e., minimizes X_K . To find this optimal combination, the primers were sorted in descending order of their discriminating power and, at each step, chosen one after the other to minimize X_K .

GenAlEx 6.5b3 (Peakall and Smouse 2012) was used to assess the genetic diversity per marker by calculating allelic richness (number of alleles and number of private alleles) and observed (Ho) and expected heterozygosity (He). In GenAlEx, the genetic structure was studied by determining the number of private alleles for *C. annuum*, *C. baccatum*, *C. chinense* and *C. frutescens*.



The previous grouping analyses were complemented with Bayesian model-based clustering using the software STRUCTURE 2.3.4 (Pritchard et al. 2000), which differentiates groups of genotypes that have distinctive allele frequencies without a priori groupings and estimates the most likely number of genetic groups (K). STRUCTURE was run for K values ranging from 1 to 12. Each run was performed using the admixture model with 10,000 replicates for burn-in and 100,000 during the analysis (Falush et al. 2007; Pritchard et al. 2000), and 10 simulations per K value. The STRUCTURE output was summarized using STRUCTURE HARVESTER (Earl and von Holdt 2012). Both, the Evanno et al. (2005) delta K test and the method of Pritchard et al. (2000) [K with the highest Pr (X|K)] were performed to estimate the optimal value of K. CLUMPAK (Kopelman et al. 2015) was used for the graphical presentation of the STRUCTURE results.

Results

Molecular variation detected with the SSR markers

Table 3 presents the description of the 21 loci based on the number of alleles detected. A total of 208 alleles



were identified, giving an average of 9.9 alleles per locus. The number of alleles ranged from five (markers NG20 and BD70) to 16 (marker NG17). The smallest allele (approx. 82 bp) was detected by SSR BD22 and the largest allele (approx. 800 bp) by NG15.

Alleles observed in only one or two of the 42 genotypes studied (frequency < 5%) were considered rare alleles (Table 3). Overall, 102 rare alleles were detected by 20 of the 21 markers evaluated. For each locus, the number of alleles and the number of rare alleles showed a significant positive correlation (r = 0.90; P < 0.0000). The highest percentages of rare alleles were detected by markers NG6 (78%) and NG7 (73%).

The absence of an amplification product for any given genotype-marker combination indicated the presence of genotypes with null alleles in the given locus. Null alleles are alleles that are not amplified

during PCR, perhaps due to polymorphism at the hybridization sites of one or both primers (Dakin and Avis 2004). Genotype-marker combinations that exhibited null alleles were replicated at least twice to ensure that absence of amplification product was not due to experimental errors.

Thirteen of the 21 loci had null alleles, ranging from two (5%) to 10 (24%) genotypes exhibiting null alleles (Table 3). The two wild genotypes of *C. rhomboideum* (Dunal) Kuntze exhibited null alleles in all of these 13 SSR. In seven (54%) out of these 13 loci, null alleles were detected only for the two *C. rhomboideum* genotypes. Despite the known prevalence of null alleles, their evolutionary dynamics and model of variation in populations is still debated (Putman and Carbone 2014); as a result, little is known of their true impact on the estimation of population differences.

Table 3 Description of the 21 microsatellites evaluated using the 42 genotypes of *Capsicum* spp

SSR	Alleles	Allele s	size (bp)	Rare	Patterns	Genotypes w	ith:	Common allele	
		Min.	Max.	alleles	observed	Null alleles	Multiple alleles	Size (bp)	Freq. (%)
NG4	10	89	114	4	15	0	6	98	29.8
NG5	7	300	324	3	8	3	2	312	38.5
NG6	9	183	201	7	9	0	0	192	42.9
NG7	15	143	193	11	15	0	0	176	33.3
NG8	9	186	218	2	10	2	1	197	25.0
NG9	6	242	327	2	7	0	1	297	51.2
NG10	9	236	277	4	10	2	1	245	38.8
NG12	14	128	190	10	14	2	5	136	32.5
NG15	9	338	800	4	12	0	4	355	27.4
NG17	16	200	422	10	17	0	1	248	16.7
NG18	14	123	146	7	14	5	0	132	18.9
NG20	5	196	217	2	5	2	0	212	65.0
NG33	9	136	154	4	9	2	0	149	37.5
NG34	9	106	162	4	9	11	0	142	32.3
NG35	7	226	250	3	7	3	0	242	43.6
BD12	11	268	344	3	12	2	1	296, 300, 307	15.0
BD17	12	196	235	6	12	0	0	203	26.2
BD22	12	82	104	6	13	7	2	94	22.9
BD33	8	110	125	3	8	3	0	125	33.3
BD41	12	190	231	7	11	10	1	208	28.1
BD70	5	150	165	0	7	0	2	154	32.1
Total	208	n.a.	n.a.	102	224	n.a.	n.a.	n.a.	n.a.
Average	9.9	n.a.	n.a.	4.9	10.7	2.6	1.3	n.a.	32.9



Any genotype-marker combination that produced two groups of bands was classified as heterogeneous (a mixture of alleles or multiple alleles). Genotypes with multiple alleles were detected by 62% of the 21 loci (Table 3). On average, 1.6 genotypes per locus presented multiple alleles. Markers NG4 (6 genotypes), NG7 (6), and NG12 (5) had the highest number of genotypes with multiple alleles.

There was a considerable variation in the frequency of the most common allele (Table 3). Among markers, a significant negative correlation was observed between the number of alleles and the frequency of the most common allele (r = -0.66, P = 0.0011). On average, at any given locus, the common allele was shared by 33% of the genotypes evaluated.

The level of polymorphism among the 42 genotypes was evaluated by calculating a PIC value for each locus (Table 4). PIC values ranged between 0.5004 (NG20) and 0.8988 (NG17), with an average of 0.7748 per marker. Based on PIC values, 20 of the SSRs can be considered informative (PIC > 0.50), four of these being highly informative (PIC > 0.85).

PIC values showed significant positive correlation with the number of alleles (r = 0.79; P < 0.0000), significant negative correlation with the frequency of the most common allele (r = -0.95; P < 0.0000) and a non-significant very low correlation with the range of allele size (r = 0.17; P = 0.4627), which indicates that microsatellites detecting the highest number of alleles were the most informative, regardless of differences in allele size revealed.

Discriminatory power of the SSR markers

As seen with the PIC value, the estimate of the discriminating power (Dj) is based on the allele frequencies. Thus, they both correspond to nearly identical values (Tessier et al. 1999), and this was observed in the present study (Table 3 and 4). In fact, the PIC values and D estimations showed a significant high positive correlation (r = 0.98; P < 0.0000), which suggests that these two estimations give highly similar information for the SSR. Nonetheless, the discriminating power reveals that the efficiency of a

Table 4 Estimation of the informativeness of each SSR on the sample of 42 *Capsicum* genotypes

SSR	PIC	Не	Но	Confusion probability		Discrimina	nting Power
				$\overline{\mathrm{C}_{j}}$	С	D	D_L
NG17	0.8988	0.9062	0.024	0.0697	0.0918	0.9303	0.9082
NG18	0.8986	0.9058	0.000	0.0691	0.0942	0.9309	0.9058
BD12	0.8690	0.8809	0.025	0.0923	0.1150	0.9077	0.8850
BD22	0.8520	0.8657	0.057	0.0958	0.1216	0.9042	0.8784
BD17	0.8402	0.8549	0.000	0.1243	0.1451	0.8757	0.8549
NG8	0.8387	0.8547	0.025	0.1205	0.1425	0.8795	0.8575
BD41	0.8211	0.8384	0.031	0.1351	0.1621	0.8649	0.8379
NG7	0.8174	0.8322	0.000	0.1475	0.1678	0.8525	0.8322
NG4	0.7978	0.8200	0.143	0.1243	0.1451	0.8757	0.8549
NG12	0.7919	0.8128	0.125	0.1615	0.1825	0.8385	0.8175
NG15	0.7866	0.8112	0.095	0.1336	0.1531	0.8664	0.8469
NG34	0.7814	0.8054	0.000	0.1677	0.1946	0.8323	0.8054
NG33	0.7774	0.7975	0.000	0.1821	0.2025	0.8179	0.7975
BD33	0.7723	0.7982	0.000	0.1808	0.2018	0.8192	0.7982
NG10	0.7613	0.7841	0.025	0.1846	0.2050	0.8154	0.7950
NG5	0.7265	0.7597	0.051	0.2119	0.2321	0.7881	0.7679
BD70	0.7146	0.7557	0.048	0.2056	0.2245	0.7944	0.7755
NG6	0.6961	0.6961	0.000	0.2869	0.3039	0.7131	0.6961
NG35	0.6896	0.7258	0.000	0.2551	0.2742	0.7449	0.7258
NG9	0.6384	0.6743	0.024	0.2938	0.3107	0.7062	0.6893
NG20	0.5004	0.5363	0.000	0.4500	0.4638	0.5500	0.5363

The markers are sorted based on descending discriminating power He, expected heterozygosity; Ho, observed heterozygosity; C_i , confusion probability for the jth SSR; C, confusion probability for the jth SSR as N tends toward infinity; D, discriminating power calculated $(1 - C_i)$; D_L , discriminating power estimated from P_i^2 (1 – C); Pi, frequency of the ith pattern of a given SSR



given microsatellite marker does not depend only on the number of patterns it generates (Table 4). For instance, two SSRs producing a similar number of patterns can have very different discriminating powers, as illustrated by NG9 and BD33. In contrast, two markers that differ significantly with respect to the number of patterns observed may have similar discriminating powers (e.g. NG12 and NG34; 14 and 9 patterns, respectively). According to Tessier et al. (1999), the differences among the frequencies of the patterns generated by the markers help explain this result. When a marker generates patterns at the same frequency, then its discriminating power is maximal. Conversely, the farther the SSR is from this situation, the more its discriminating power diminishes.

Based on the values of the estimated total number of non-differentiated pairs, one optimal combination of three SSR markers would be sufficient to discriminate among the 42 genotypes studied (Table 5). In the present study, these three SSR markers, generated from the primers, NG17 + NG18 + BD12, were selected on the basis of their discriminating power (Table 4).

Resolving power of the SSR markers by cluster analysis

The 21 microsatellites evaluated made it possible to distinguish among the studied 42 genotypes. Figure 1 shows the UPGMA clustering pattern for the 42 genotypes with two main clusters. One cluster was formed by the two genotypes representing the wild species *C. rhomboideum* (Group A). These two genotypes showed virtually no similarity with the rest of the genotypes. This species presents a combination of morphological characteristics that contrasts with the other species of the *Capsicum* genus. The flowers of *C. rhomboideum* are yellow, unlike the other species that have white or purple flowers, and it is considered as a non-pungent and basal species of the *Capsicum* genus (Carrizo García et al. 2016; Egan et al. 2019; Walsh and Hoot 2001).

The other main cluster grouped together genotypes from both domesticated and wild species, hampering a conclusive interpretation based on taxonomic status. Nonetheless, based on bootstrap values greater than or equal to 75%, it can be interpreted that, within this main cluster, there are six minor clusters that correspond to: (B) two genotypes of *C. pubescens* and one

of *C. minutiflorum* Rusby (Hunz.); (C) two genotypes of *C. chacoense* Hunz.; (D) eleven genotypes of *C. annuum* (including the one from *C. annuum* var. *abbreviatum* Fingerh.) and one genotype of each species, *C. baccatum*, *chinense x annuum*, and *C. luteum* Lam.; (E) three genotypes of *C. chinense* and three of *C. frutescens*; (F) seven genotypes of *C. baccatum* and one of *C. frutescens*; and (G) four genotypes of *C. chinense*. Three genotypes did not belong to any of these groups but are still different from each other, making up unitary groups. These genotypes represented the domesticated species *C. frutescens* (16fru), and the wild species, *C. cardenasii* (24car) and *C. flexuosum* Sendtn. (25flex).

The cluster procedure applied to the coordinates obtained by PCoA revealed that the first three axes accounted together for 53.56% of the total variation, with 25.40, 15.54, and 12.62% explained, respectively, by PC axis 1, 2 and 3 (data not shown). The representation of the distribution of the genotypes in a tri-dimensional PCoA space suggested a structure of four main clusters (Fig. 2), corresponding to one group that included the 10 C. annuum genotypes (including var. abbreviatum) plus the chinense × annuum and C. luteum genotypes; a second group clustering three C. chinense genotypes and three of C. frutescens; a third group formed by six C. baccatum genotypes and one C. frutescens; and a fourth group composed of 14 genotypes representing species C. cardenasii, C. chacoense, C. flexuosum, C. pubescens and C. rhomboideum. Consistently in our clustering analyses, one C. baccatum genotype showed no evident close relation to other genotypes based on taxonomic species.

In our results from Bayesian model-based clustering, the delta K statistic in the Evanno test and median values of $\ln(\Pr Data)$ showed that K=10 was the optimal number of genetic groups in this study (Fig. 3). For K=10, the clustering in different runs was almost identical (similarity coefficient 0.890). The run with the highest likelihood value ($\ln L=-1216.5$) was selected to classify genotypes based on their posterior assignment probabilities (more than 50%) and to explore the possibility of a taxonomical explanation of these K=10 genetic clusters (Table 1). These clusters mainly merged genotypes from the same species, thus a pattern of taxonomic structure can be deduced. For instance, clusters C1, C2, C3 and C4 grouped together genotypes of C.



Table 5 Number of non-differentiated pairs obtained with several primer combinations

SSR	Order of D	Indistinguishable pairs per SSR	Primer combination	Indistinguishable pairs per combination
NG17	1	60	NG17	60
NG18	2	59	NG17 + NG18	4
BD12	3	79	NG17 + NG18 + BD12	0
BD22	4	82	NG17 + NG18 + BD12 + BD22	0
BD17	5	107	NG17 + NG18 + BD12 + BD22 + BD17	0
NG8	6	104	NG17 + NG18 + BD12 + BD22 + BD17 + NG8	0
BD41	7	116	NG17 + NG18 + BD12 + BD22 + BD17 + NG8 + BD41	0
NG7	8	127	NG17 + NG18 + BD12 + BD22 + BD17 + NG8 + BD41 + NG7	0
NG4	9	107	-	0
NG12	10	139	-	0
NG15	11	115	-	0
NG34	12	144	-	0
NG33	13	157	-	0
BD33	14	156	-	0
NG10	15	159	-	0
NG5	16	182	-	0
BD70	17	177	-	0
NG6	18	247	-	0
NG35	19	220	-	0
NG9	20	253	-	0
NG20	21	387	-	0

annuum, C. chinense, C. frutescens and C. baccatum, respectively; these species were represented by at least five genotypes.

The split using a model-based Bayesian analysis was congruent with UPGMA clustering and PCoA analysis, although it was more discriminating in that it included nine of 10 genotypes from C. annuum in C1 (the other one clustered together with annuum × chinense, var. abbreviatum and C. luteum in C6) and four of seven genotypes from C. chinense in C2; whereas C. baccatum was contained in C4 and C. pubescens in C5, and still depicted an intermixed cluster (C3) of *C. frutescens* and *C. chinense* (Table 1; Fig. 3). This split also clearly grouped together the two genotypes of wild species C. rhomboideum (Cluster C10). In turn, C. minutiflorum (C8) and C. chacoense (C9) stood out in one cluster per species, and C. flexuosum and C. cardenasii made-up an admixed cluster (C7). One genotype of C. baccatum (20bac) had a posterior assignment probability value of less than 50%, which seems to be indicative of an admixed genotype (with genetic contributions from more than one cluster).

Markers to distinguish among same species genotypes

Table 6 summarizes the SSR description for species with five or more genotypes included in this evaluation. Due to their polymorphism, 15 SSRs were useful to differentiate the 10 genotypes of *C. annuum* (Table 6). None of these markers detected null alleles in any of these genotypes. The genotype *C. annuum* var. Chung yang showed multiple alleles in loci NG4, NG15, BD22 and BD70 (data not shown). The genotypes Jalapeño 3575, Jalapeño 1493, and Papri-Queen exhibited multiple alleles in markers NG4, NG9, and NG15, respectively (data not shown).

Twenty microsatellites were polymorphic within the genotypes of *C. baccatum* (Table 6). PIC values



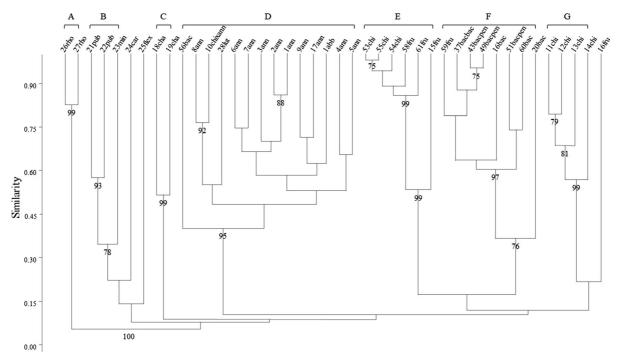
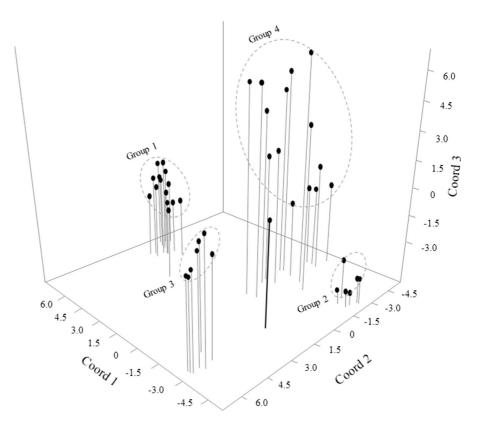


Fig. 1 UPGMA dendrogram, based on Nei's genetic distance, depicting the genetic relationships among the 42 genotypes studied. Numbers at the nodes provide branch support $\geq 75\%$

based on 10,000 bootstrap replicates. Capital letters indicate the different genetic groups

Fig. 2 Spatial distribution of the groups of genotypes obtained with the three main axes of variation by PCoA





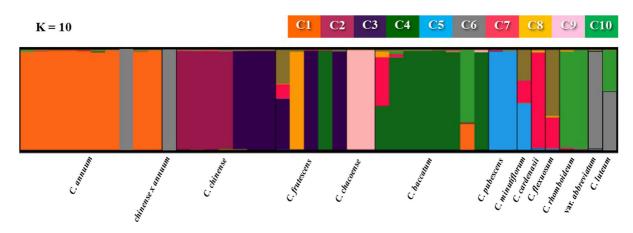


Fig. 3 Genetic groups formed by the 42 studied genotypes at K = 10

Table 6 Number of alleles and polymorphism information content (PIC) found in the 21 microsatellites within the genotypes of *C. annuum*, *C. baccatum*, *C. chinense* and *C. frutescens*

SSR	C. annuum $n = 10$		C. bacca	<i>tum</i> n = 8	C. chiner	<i>ise</i> n = 7	C. frutescens $n = 5$	
	Alleles	PIC	Alleles	PIC	Alleles	PIC	Alleles	PIC
NG4	3	0.442	2	0.195	3	0.523	3	0.499
NG5	1	0.000	3	0.605	2	0.370	4	0.610
NG6	1	0.000	2	0.219	2	0.490	2	0.320
NG7	3	0.314	5	0.712	5	0.700	2	0.269
NG8	4	0.596	2	0.195	2	0.325	5	0.768
NG9	2	0.351	1	0.000	2	0.370	2	0.269
NG10	4	0.450	3	0.371	2	0.370	2	0.269
NG12	1	0.000	4	0.547	1	0.000	4	0.672
NG15	3	0.572	5	0.618	2	0.370	3	0.499
NG17	2	0.375	4	0.630	3	0.530	3	0.672
NG18	4	0.692	4	0.605	3	0.530	4	0.672
NG20	1	0.000	2	0.195	3	0.530	3	0.499
NG33	3	0.548	4	0.524	4	0.641	2	0.269
NG34	5	0.642	4	0.712	3	0.641	2	0.499
NG35	1	0.000	2	0.195	3	0.406	3	0.563
BD12	2	0.375	4	0.483	3	0.530	3	0.499
BD17	2	0.164	3	0.555	2	0.370	3	0.563
BD22	5	0.716	4	0.510	3	0.530	4	0.672
BD33	1	0.000	2	0.195	2	0.370	3	0.672
BD41	3	0.410	2	0.371	2	0.215	2	0.499
BD70	3	0.347	2	0.195	2	0.370	3	0.499
Total	54		64		54		62	
Average	2.57	0.333	3.05	0.411	2.57	0.437	2.95	0.512

n, number of genotypes per species

indicate that loci NG7 and NG34 were the most informative markers in genotypes of this species. The SSR markers NG5 (genotype 51bacpen), NG34 (16bac and 60bac) and BD41 (16bac, 37bacbac, 43bacpen, 49bacpen, 51bacpen and 60bac) revealed

nine genotypes with null alleles. Multiple alleles were detected by markers NG12, BD12 and BD22 in genotype *C. baccatum* var. *praetermissum* (Heiser and P.G.Sm.) Hunz., and by locus NG15 in genotype *56bac*.



The seven genotypes of *C. chinense* were polymorphic when assayed with 20 microsatellite markers (Table 6). PIC values indicate that markers NG7, NG33 and NG34 were the most informative. Marker NG34 detected null alleles in genotype *14chi*. Multiple alleles were determined at locus NG4 in genotypes *12chi* and *53chi*.

All 21 microsatellites evaluated were polymorphic in the set of five genotypes representing *C. frutescens* (Table 6). PIC values show that marker NG8 was the most informative microsatellite, as this SSR detected one different allele per every *C. frutescens* genotype evaluated. In turn, microsatellites NG17, NG18, BD22 and BD33 were the second most informative markers, each informative to an identical degree. Markers NG34 (genotypes 58fru, 59fru and 61fru) and BD41 (59fru) detected four genotypes with null alleles. One genotype with multiple alleles was detected by SSR NG5 (15fru).

Discussion

Polymorphism and usefulness of the studied markers

Microsatellite markers present several advantages over other types of molecular markers, because they are considered to be robust, more variable and informative than RFLP, RAPD and AFLP (He et al. 2003; Lee et al. 2004; Senan et al. 2014). In this study, the high level of polymorphism detected among the 42 genotypes assayed demonstrated the efficiency of SSR markers for discriminating both distinct taxa and variation within a species (Table 3).

The 21 microsatellite markers used in this study generated 208 alleles, with an average of 9.9 alleles per locus. The significant positive correlation (r = 0.79; P < 0.0000) between PIC value and number of alleles per SSR, discriminating power and number of alleles (r = 0.72; P < 0.0002), as well as the positive correlation between PIC values and discriminating power (r = 0.98; P < 0.0000), suggests that markers that detect higher number of alleles are more informative and therefore more useful in genotyping and diversity analysis of *Capsicum* germplasm. Our results confirm that any of these three estimators are potentially helpful in determining

the informative value of an SSR marker used in diversity studies of the genus.

Botstein et al. (1980) reported that PIC values above 0.5 are consistent with usefulness of the marker. Based on PIC values, the level of polymorphism of the 21 microsatellites was very high, ranging from 0.500 to 0.899, with an average of 0.775. The polymorphism of these microsatellite markers could accordingly be used to discover duplicated entries in ex situ germplasm collections of Capsicum, as well as to resolve disputes related to intellectual property rights related to Capsicum germplasm. Furthermore, this high level of information was used to discriminate the genotypes in the present study, as shown by UPGMA clustering, PCoA analysis and Bayesian model-based clustering. The results from this study clearly confirm that microsatellites are a promising approach to definitive differentiation among cultivated varieties or different species of the genus Capsicum.

Rare alleles are highly informative for genotyping of varieties (Jain et al. 2004) and can serve as an indicator of the presence of unique genetic variants or differentiated populations. Genetic erosion, on the other hand, can rapidly lead to the loss of these alleles (Bijlsma and Loeschcke 2012). Nonetheless, this susceptibility makes them valuable tools for assessing losses of genetic diversity that occur during multiplication and regeneration of germplasm collections. The 20 microsatellites markers that detected rare alleles in this study are likely candidates for broad use within the genus for cultivar discrimination or identification of species of the genus Capsicum. They will also be helpful as baseline information in future decisions related to the most efficient way to disseminate and maintain ex situ collections of Capsicum germplasm.

Thirteen of the evaluated microsatellites detected genotypes with multiple alleles (Table 3). This level of polymorphism could be (1) the result of remnant heterozygosity in several genotypes, or (2) the consequence of the heterogeneity produced by involuntary mixture of seed. In turn, the residual heterozygosity could be the result of allogamy and would be a foreseeable characteristic in materials that have not been bred to develop pure lines. *Capsicum* is regarded as a primarily self-pollinating crop (Allard 1971); however, allogamy rates in *Capsicum* suggest that it should be considered as a facultative cross-pollination plant (Tanksley 1984). In the present study, heterozygosity cannot be clearly distinguished from



heterogeneity because leaf tissue from three plants per genotype was pooled to extract the DNA. Nonetheless, the presence of allelic variants is highly valuable to find distinctive markers useful for discrimination of a given genotype.

Most of genotypes evaluated in this study (genotypes 1-30 in Table 1) have been multiplied by selfpollination under greenhouse conditions at Cornell University (former Jahn laboratory working group) over a period of several years to obtain seeds that are used in breeding programs. Although involuntary mistakes during handling cannot be discarded, it can be assumed that there was a very low likelihood of seed mixture events during the management of these genotypes. Therefore, genotypes provided by Cornell University that exhibited multiple alleles (e.g., Jalapeño 3575, Jalapeño 1493, PapriQueen and var. praetermissum) reinforce that the 21 microsatellites evaluated in this study are helpful markers to discover new molecular variability, or allelic variants, that can serve as distinctive markers for a given genotype.

The remaining genotypes were provided by the USDA (genotypes 31–42 in Table 1), but it is unclear the way these materials were managed during their conservation. Anecdotal reports suggest that any accession in the collection in the early 1980s was put through a round of open pollination in field increases early in that decade (Jahn, personal communication). This means that multiple alleles exhibited by genotypes of this subset (e.g. 53chi and 56bac in markers NG4 and NG15, respectively) may be attributable to allogamy in Capsicum. On the other hand, due to the lack of sound information about their management, it is also probably that these USDA genotypes showing multiple alleles reflect either unintentional mixture of seed that may have occurred during germplasm management, or inadvertent mistakes taken place while assigning introduction numbers (PI) in germplasm banks from which the genotypes were originally obtained. Accordingly, this study also identifies markers such as NG4 and NG15, that may be used by genebanks as an effective tool to detect mishandling that could have occurred during regeneration, multiplication and management of Capsicum germplasm.

Carrizo García et al. (2016) presented the most inclusive molecular phylogeny of *Capsicum* published to date using most of the recognized species of the genus. This study showed 11 well-supported clades,

however the placements of some species and their closest relatives are not strongly resolved. In the present study, the Bayesian model-based analysis performed using software STRUCTURE to estimate the most probable K number of genetic groups, produced a most probable value of K = 10. This configuration of the 42 studied genotypes corresponded to some degree to the species included in the study. C. rhomboideum genotypes separate from the rest of the species (Cluster C10), as expected according to the classification indicated by cytogenetic and molecular data (Carrizo García et al. 2016; Moscone et al. 2007). According to the phylogenic analysis by Carrizo García et al. (2016), C. rhomboideum belongs to the Andean clade, the most ancient group of the genus which includes nonpungent species native to the Andes, western to north-western South and Central America. Further studies have indicated that this species is sister to the rest of the genus (Guzmán et al. 2009; Särkinen et al. 2013; Walsh and Hoot 2001).

As for the five domesticated species, the clustering of C. annuum genotypes together with one C. chinense genotype (C1) and also three genotypes of C. frutescens grouped with C. chinense (C3) reflects the strong relationship among these three species (Carrizo García et al. 2016). Together, these two clusters can be understood because these species form the C. annuum complex (C. annuum, C. chinense and C. frutescens). The difficulties in taxonomic differentiation within these three taxa have been previously reported (Baral and Bosland 2004; Moscone et al. 2007; Walsh and Hoot 2001), consistent with their high degree of genetic uniformity. In turn, C. baccatum genotypes are grouped in cluster C4. This species is characterized by white flowers, similar to those in the C. annuum complex (Carrizo García et al. 2016; Scaldaferro et al. 2018). It is remarkable that cluster C5 was composed of C. pubescens, the most distinct of any of the five domesticated species; characterized by strongly violet corollas, blackish brown seeds and pubescence on the entire plant (Carrizo García et al. 2016).

Owing to their codominance, multi allelism, and cross-species amplification, microsatellites are one of the most extensively harnessed markers in several areas of plant research (Senan et al. 2014). Microsatellite polymorphisms can be detected by agarose or polyacrylamide gel electrophoresis, and SSR alleles can be visualized using ethidium bromide or silver



staining. Their reproducibility allows compare data among laboratories and their protocols imply the use of relative low-cost reagents and equipment. As a result, microsatellites can be very helpful in developing countries or any context where limited funding may constrain the scientific community working toward long term conservation of genetic diversity within *Capsicum* and use of *Capsicum* genetic resources.

With the aim to contribute to safeguarding ex situ collections of *Capsicum* spp., the molecular resources presented in this study define a major advance for the conservation of *Capsicum* genetic resources, especially those from developing countries, allowing much more accurate and efficient assessment of *Capsicum* spp. germplasm to support better decisions in day-to-day conservation activities with highly constrained resources.

Enabling conservation and use of *Capsicum* genetic diversity

The conservation of genetic resources of *Capsicum* sp. in germplasm banks is justified by the need and ongoing, even intensified use of this germplasm by both public and private sector plant breeders. In turn, use of *Capsicum* germplasm among researchers and conservation activities requires reliable information about the genetic makeup of the conserved accessions. Thus, ex situ collections are fundamental for the present and future utilization of *Capsicum* sp. genetic resources.

This study characterizes a set of 21 polymorphic microsatellite markers that successfully amplify across 11 Capsicum species, establishing their utility for broad molecular characterization of genetic diversity of Capsicum species. These results spotlights the relevance of this set of markers for both enhanced characterization of Capsicum germplasm and potentially improved efficiencies in breeding programs. Due to their polymorphism, these microsatellites can further be used to estimate changes in the genetic structure of ex situ collections of Capsicum germplasm. In addition, available information on allelic frequencies can serve as a baseline to make sound decisions about conservation of *Capsicum* germplasm. Allelic variants that occur at low frequency can be identified and maintained, and sources of genetic variability can be ascertained through genotyping of local varieties. Finally, these markers can also be used to trace the recent genetic history of a given modern commercial variety or landrace endemic to a geographical region.

The use of these microsatellites to characterize *Capsicum* germplasm conserved ex situ will promote the utilization of more collections which are the foundation for the development of improved varieties. These markers will also be helpful in assessing to what extent any *Capsicum* ex situ collection is representative of the genetic diversity present in a specific country or region, and in identifying gaps within collections, thus providing guidelines to define complementary conservation strategies.

The overall results of this study highlight that identification and establishment of useful, efficient microsatellites can offer consistent results and enable preservation of genetic distinctness, uniformity, and discrimination of promising genotypes for *Capsicum* breeding. Morphological markers, in conjunction with molecular genetic evidence provided by this set of microsatellites, would be beneficial for definition of strategies for complementary conservation of *Capsicum*, as well as for selection of germplasm to be included in breeding programs.

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Compliance with ethical standard

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

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