

C a s f c sa e es a d SNPs f ed g ee a a s s e Pac f c s e Crassostrea gigas

T g L 1 · Q L 1,2 · L gfe g K g 1 · H g Y 1

Received: 3 August 2016 / Accepted: 9 February 2017 © Springer International Publishing Switzerland 2017

Abs ac Knowledge of the pedigree relationships between individuals is a prerequisite in genetics research, and the application of molecular markers for pedigree analysis has been a booming science for over a decade. Owing to the high variability, microsatellites are considered as the marker of choice for studies on pedigree analysis. Nevertheless, single nucleotide polymorphisms (SNPs) have been increasingly used for this purpose in recent years due to the low mutation rate and genotyping error rate. To compare the utility of microsatellites and SNPs in assigning parentage in the Pacific oyster (Cra rea *** a**), we genotyped 384 parental and offspring individuals using 12 multiplexed microsatellites and 50 SNPs. In this study, all microsatellite loci showed high informative (PIC >0.5), while most SNPs were middle informative (0.25 < PIC < 0.5). CERVUS simulations revealed that using nine microsatellites or 38 SNPs, the power of parental assignment could reach 100%. Pedigree analysis of real offspring demonstrated that 100% of the offspring were unambiguously assigned to a pair of parents when nine microsatellites or 50 SNPs were used. For microsatellites, the combined exclusion power with one parent known (EXCL2) could reach one when three microsatellites multiplex PCRs or more were used, whereas EXCL2 was 0.9999 for the 50 SNPs. In general, six SNPs were needed to obtain an equivalent exclusion power for pedigree analysis with a microsatellite locus in C. ** . The information obtained in this study will be useful for assigning parentage in C. ** using both marker systems.

Ke d ◆ *Cra* rea • Microsatellite multiplex PCR · Single nucleotide polymorphisms · Pedigree analysis

I dc

Aquaculture is now the fastest growing animal production worldwide. In Asia Pacific region, it significantly contributes to local food/nutritional security (FAO 2014). Such an important industry is expected to be based on genetically improved stocks. As a result, genetic improvement of aquaculture stocks is urgently needed. However, genetic improvement is still limited due to the complexity and cost of maintaining adequate pedigree information in aquaculture stocks. A pedigree is one of the simplest concepts in biology (Pemberton 2008), but it is of paramount importance for breeding program as well as genetic management to obtain sustainable genetic improvement in aquaculture industry. Correct pedigree information is necessary for the accurate estimation of breeding values, whereas inaccurate and incomplete genealogies lead to deviations and errors in breeding program and genetic management. Therefore, the provision of a credible and cost-effective method to obtain pedigrees in any species and rearing system was supposed to be of great interest for a successful breeding program in aquaculture.

DNA markers are becoming increasingly important in animal breeding. Molecular parentage analysis using DNA markers has made the pedigree tracing possible in aquaculture species. This method is expected to be promising as it can avoid the initial investment in separate family rearing units and limits associated biases, even more in species with high larval mortality, small larval size, and initial live feeding (Vandeputte and Haffray 2014).

The principle of molecular pedigree analysis is very simple. It is based on the simple concept that parents pass on one of two alleles at each locus to the offspring, which therefore carries one allele from each parent (Herbinger et al. 1995). Microsatellites are of high variability and wide availability. Thus, they have been the preferred molecular markers to trace the genealogical relationships in numerous aquaculture species (Castro et al. 2007; Lallias et al. 2010; An et al. 2011; Vandeputte et al. 2011; Fu et al. 2013; Zhang et al. 2016). However, considering the presence of null alleles and higher mutation rate, which would interfere with accurate pedigree analysis with microsatellites, application of single nucleotide polymorphisms (SNPs) is growing exponentially (Guichoux et al. 2011) and they have been applied for large-scale parentage studies (Hauser et al. 2011; Pino-Querido et al. 2015) because of the speed of high-through-put screening, low genotyping error and easy transfer-ability between laboratories (Anderson and Garza 2006). In addition, SNPs have lower mutation rates (10⁻⁹) per locus per generation than microsatellites $(10^{-3}-10^{-4})$ (Ellegren 2000), and they occur more frequently in the genome than microsatellites (Bester et al. 2008). Nevertheless, the resolving power of SNPs for pedigree assignment needs be compensated by large marker sets because each SNP locus typically has only two alleles.

As one of the most commercially important species in aquaculture, the Pacific oyster (Cra rea in) has the highest worldwide production of cultured aquatic species. In China, C. is still in an early stage of domestication. The marker-assisted selection could accelerate the development of the oyster industry and maximize efficiency of aquaculture production. A large number of microsatellite and SNP markers have been developed in C. in (Sekino et al. 2003; Yamtich et al. 2005; Bai et al. 2009; Zhong et al. 2013; Lapegue et al. 2014). Initially, microsatellite was amplified as single locus for pedigree analysis. To reduce the cost and time required for microsatellite genotyping so that shed the burden of laboratory works, multiplex PCR panels of microsatellites of C. in have been developed (Taris et al. 2005; Li et al. 2010; An et al. 2013; Liu et al. 2016).



In the present study, in order to compare the utility of microsatellites and SNPs in assigning parentage in *C.* *** we genotyped more than 300 individual oysters using 12 multiplexed microsatellites and 50 SNPs and examined the results of parentage analyses using both marker systems and the resolving power of analyses employing different numbers of microsatellites and SNPs.

Maeasad e ds

E e e a sa es a d DNA e ac

In 2007, 2-year-old Pacific oysters from one base population in Rushan, Shandong province, were used to establish the first-generation selection line for fast growth. The oysters derived from natural seed which were collected on local coast, and were cultured on ropes suspended from rafts along the coastal regions. Samples of one-year-old *C.* ** were collected in October 2015 from one cultured population in Rongcheng, Shandong province, China. They were the eighth-generation offspring produced by successive selection. In July 2014, 100 individuals (50 males and 50 females) were selected from the seventh-generation strains to serve as parents for the eighth-generation selected strain. Tissues from all individuals were saved in pure ethanol until DNA extraction.

Genomic DNA was extracted from adductor muscle tissue as previously described by Li et al. (2006). The concentration and quality of DNA were assessed using a NanoDrop 2000 spectrophotometer and by running a small amount on a 1% agarose gel. The extracted genomic DNA was stored at -20 °C until genotyping.

To provide a benchmark for comparison, 100 parents and 284 progenies were genotyped by 12 microsatellites and 50 SNP markers, respectively. The microsatellite system consisted in a total of 12 microsatellite DNA markers multiplexed. In brief, forward primers were modified with M13-tail with different fluorescent dyes (FAM, PIT, VIC and NED). PCR reactions were carried out in 10 μ l volumes containing 1× PCR buffer, 0.2 mM dNTP mix, 2.0 mM MgCl₂, 0.15 μ M forward primer, 0.06 μ M reverse primer, 0.15 μ M universal primer, 0.25 U τ a DNA polymerase, and about 50 ng template DNA. Thermal cycling was as follows: first denaturation at 94 °C for 3 min; then 35 cycles of 30 s at 94 °C, 60 s at the optimal annealing temperature, and 75 s at 72 °C; 8 cycles of 30 s at 94 °C, 60 s at 53 °C, 75 s at 72 °C, with a final extension of 10 min at 72 °C. PCR products were genotyped on ABI-3130 with LIZ500 as internal size standard. Data were processed with the GeneMapper v4.0 software.

For the 50 previously developed SNPs (Zhong et al. 2013), PCR was performed in a 10 μl reaction mixture on a LightCycler® 480 real-time PCR instrument (Roche Diagnostics). The mixture contained 10× PCR buffer, 0.2 mM dNTP mix, 1.5 mM MgCl₂, 0.2 μM of each primer set, 5 μM SYTO®9 (Invitrogen Foster City, CA, USA), 0.25 U *Ta* DNA polymerase, and about 10 ng template DNA. The amplification procedure was achieved as follows: initial denaturation at 95 °C for 5 min, 45–50 cycles of denaturation at 95 °C for 10 s, annealing temperatures at 60 °C for 10 s, then extension at 72 °C for 10 s. Following amplification, melting curves were generated by collecting fluorescence data between 60 °C to 90 °C. Data were analyzed using the LightCycler® 480 Gene Scanning Software 1.5 (Roche Diagnostics).



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We calculated the number of alleles (Na), polymorphic information content (PIC), the observed heterozygosity (H), and expected heterozygosity (He) using CERVUS 3.0 (Kalinowski et al. 2007) for each microsatellite locus. All microsatellites were tested for Hardy–Weinberg equilibrium with a chi-square test (using Yates' correction) in CERVUS 3.0, and those loci showed with "not done" were tested again using exact test implemented in the program GENEPOP v 4.5.1 (Rousset 2008). The frequency of null alleles at each locus from population data was obtained following the method by Van Oosterhout et al. (2004) as implemented in the Micro-Checker 2.2.3 software. Basic genetic parameters of SNPs including Na, PIC, H, He, HWE, and null allele frequency were computed as for microsatellites. Allele frequencies for microsatellites and SNPs in 384 individuals were calculated directly.

Pedigree analysis of microsatellite and SNP genetic data was consequently performed with the likelihood-based approach in CERVUS 3.0. Both simulation and real parentage analysis were conducted. The exclusion probability of each locus based on the genotype of no parent known and on the genotype of one parent known was named as EXCL1 and EXCL2, respectively. Both EXCL1 and EXCL2 were calculated for each locus using program CERVUS 3.0. And the combined exclusion probabilities of different marker sets were also calculated. The exclusion power ($P_{\rm u}$) of microsatellites multiplex PCRs and SNPs was also computed using Eqs. 1 and 7 as described in Vandeputte (2012).

Res s Ge e c d e s f c sa e es a d SNPs

Genetic diversity parameters were estimated in the 384 individuals. Twelve microsatellite loci were arranged into four multiplex PCRs (Table 1). The allele number (*Na*) of these loci ranged

Tab e 1 Information of 4 microsatellite multiplex PCRs

Multiplex sets	Locus	Ta (°C)	Concentration of forward primer (µM)	Concentration of reverse primer (µM)
Panel 1	ucdCg-120	58	0.06	0.15
	ucdCg-198	58	0.06	0.15
	ucdCg-117	58	0.06	0.15
	FAM-M13 (-21)	53	_	0.15
Panel 2	Crgi3	58	0.06	0.15
	ucdCg-146	58	0.06	0.15
	uscCgi-210	58	0.06	0.15
	PIT-M13 (-21)	53	_	0.15
Panel 3	ucdCg-170	58	0.06	0.15
	ucdCg-156	58	0.06	0.15
	ucdCg-199	58	0.06	0.15
	VIC-M13 (-21)	53	_	0.15
Panel 6	otgfa0 408293	54	0.06	0.15
	otgfa0 0139 G12	54	0.06	0.15
	ucdCg-200	54	0.06	0.15
	NED-M13 (-21)	53	_	0.15

Ta annealing temperature



from a minimum of nine for Crgi3 and otgfa0_0139_G12 to a maximum of 16 for ucdCg-146 and ucdCg-156 with an average of 11.8 (Table 2). The polymorphism information content (PIC) and expected heterozygosity (He) ranged between 0.619 (uscCgi-210) and 0.917 (ucdCg-156), between 0.647 (uscCgi-210) and 0.924 (ucdCg-156), respectively. All loci showed high informative (PIC > 0.5) (Botstein et al. 1980). Allele frequency distributions were highly uneven. Some loci displayed rather homogeneous allelic frequencies (such as ucdcg-146 and ucdcg-156), while some others distributed unevenly (such as ucdcg-199 and uscCgi-210). These markers were an assortment of di-, tri-, and tetranucleotide repeats. Most of them exhibited nearly regular allelic series in conformity with the reported repetition motifs except ucdcg-117, ucdcg-199 and uscCgi-210 (Fig. 1).

The 50 SNPs were divided into four groups according to the PIC values from high to low, including 12, 13, 13, and 12 SNPs (Table 3). There were two alleles in each SNP locus. The PIC ranged from 0.202 to 0.375 and the He ranged from 0.229 to 0.501. Only three loci (CgSNP35, CgSNP149, and CgSNP176) exhibited slightly informative (PIC < 0.25), the remaining were middle informative (0.25 < PIC < 0.5) (Botstein et al. 1980).

These two types of markers showed different allele frequency distributions (Fig. 2). For the microsatellites, most of the alleles were found to be less than 10% frequency, while all of the SNPs showed an overall frequency of greater than 10%.

Significant departures from Hardy–Weinberg (HW) proportions due to heterozygote deficiency were detected both in the microsatellites (ucdCg-117, ucdCg-156, ucdCg-199, otgfa0_0139_G12 and ucdCg-200) and the SNPs (CgSNP188, CgSNP180, CgSNP65, CgSNP52, CgSNP27, CgSNP203, CgSNP34, CgSNP147, CgSNP187, CgSNP197, CgSNP148 and CgSNP162).

Pa e age a a ss

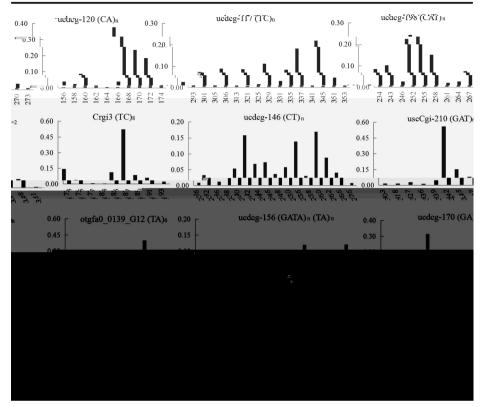
The CERVUS simulations showed that the power of parental assignment with 9 microsatellites (three multiplex panels) or 38 SNPs (three groups) could be 100% (Table 4). The actual parentage analysis demonstrated that all of the offspring were correctly allocated to a pair of

Multiplex sets	Locus	N	Na	PIC	Н	Не	EXCL1	EXCL2	HWE	F (null)
Panel 1	ucdCg-120	384	10	0.722	0.880	0.758	0.366	0.546	NS	-0.092
	ucdCg-198	384	11	0.821	0.820	0.841	0.518	0.685	NS	0.010
	ucdCg-117	384	15	0.897	0.779	0.906	0.674	0.806	*	0.069
Panel 2	Crgi3	384	9	0.655	0.875	0.683	0.294	0.481	NS	-0.254
	ucdCg-146	382	16	0.888	0.872	0.898	0.657	0.794	NS	0.013
	uscCgi-210	383	10	0.619	0.603	0.647	0.259	0.445	NS	0.010
Panel 3	ucdCg-170	383	12	0.829	0.825	0.845	0.539	0.704	NS	0.002
	ucdCg-156	383	16	0.917	0.574	0.924	0.728	0.843	**	0.183
	ucdCg-199	383	12	0.655	0.535	0.670	0.301	0.496	**	0.100
Panel 6	otgfa0_408293	383	10	0.820	0.903	0.840	0.515	0.683	NS	-0.044
	otgfa0_0139_G12	384	9	0.745	0.633	0.772	0.403	0.584	*	0.075
	ucdCg-200	383	12	0.840	0.799	0.856	0.553	0.715	*	0.026

Tab e 2 Four multiplex PCR sets and genetic parameters of 12 microsatellite loci in C.

N number of individuals, Na number of alleles, PIC polymorphic information content, H observed heterozygosity, He expected heterozygosity, EXCL1 and EXCL2, exclusion probabilities when no parent is known and when one parent is known, respectively, HWE significance of Hardy—Weinberg equilibrium test with Bonferroni correction, NS non-significant deviation from Hardy—Weinberg equilibrium, * and **, significant deviation from Hardy—Weinberg equilibrium, * Micro-checker 2.2.3





F g. 1 Allele frequency distributions of the 12 microsatellites in the 384 individuals of Cra rea 🚉

parents based on three microsatellites multiplex panels or 50 SNPs (four groups) (Table 4). For microsatellites, EXCL1 values ranged from 0.259 (uscCgi-210) to 0.728 (ucdcg-156) and EXCL2 from 0.445 (uscCgi-210) to 0.843 (ucdcg-156) (Table 2). The combined exclusion probability values, both EXCL1 and EXCL2, were higher than 0.98 when two multiplex PCR panels were used (Fig. 3). Moreover, the combined exclusion probability could reach 1 for EXCL2 when three or four panels were applied. For SNPs, EXCL1 ranged from 0.026 (CgSNP176) to 0.125 (CgSNP96, CgSNP144 and CgSNP192) and EXCL2 from 0.101 (CgSNP176) to 0.278 (CgSNP224) (Table 3). The combined exclusion potentials with no parent known (EXCL1) and one parent known (EXCL2) for the 50 SNP loci were 0.9909 and 0.9999, respectively (Fig. 4). The exclusion power ($P_{\rm u}$) was 0.9922 when three microsatellites multiplex panels or 50 SNPs were applied, while $P_{\rm u}$ reached 0.9995 when all four microsatellite multiplex panels were used (Fig. 5).

D sc ss

A variety of SNP genotyping methods have been developed in aquaculture animals, including tetra-primer ARMS-PCR, TaqMan, melting temperature (Tm)-shift, and Illumina Golden Gate technology (Bai et al. 2009; Hansen et al. 2011; Liu et al. 2011; Lapegue et al. 2014). High-resolution melting (HRM) is a very attractive, fast, and cost-effective SNP genotyping technology because it is simple, nondestructive, and amendable to high-throughput on 96-or 384-well plates (Wittwer 2009). For this technique, only polymerase chain reaction (PCR), a

Tab e 3 Four SNP groups and genetic parameters of 50 SNP loci in C.

Groups	SNP Name	N	Na	PIC	Н	Не	EXCL1	EXCL2	HWE	F (Null)
Group 1	CgSNP96	380	2	0.375	0.692	0.500	0.125	0.187	NS	-0.161
	CgSNP144	378	2	0.375	0.712	0.501	0.125	0.187	NS	-0.174
	CgSNP192	384	2	0.375	0.706	0.501	0.125	0.187	NS	-0.170
	CgSNP33	380	2	0.374	0.063	0.498	0.124	0.187	NS	-0.007
	CgSNP85	380	2	0.374	0.845	0.498	0.124	0.187	NS	-0.259
	CgSNP186	384	2	0.374	0.466	0.499	0.124	0.187	NS	0.003
	CgSNP230	384	2	0.373	0.430	0.496	0.123	0.186	NS	0.007
	CgSNP131	382	2	0.371	0.654	0.492	0.121	0.185	NS	-0.142
	CgSNP224	384	2	0.370	0.352	0.491	0.120	0.278	NS	-0.165
	CgSNP158	380	2	0.369	0.511	0.489	0.119	0.185	NS	-0.022
	CgSNP188	384	2	0.367	0.352	0.484	0.117	0.183	**	0.158
	CgSNP04	373	2	0.365	0.558	0.481	0.116	0.183	NS	-0.073
Group 2	CgSNP14	378	2	0.365	0.429	0.480	0.115	0.182	NS	0.016
	CgSNP157	381	2	0.364	0.459	0.480	0.115	0.182	NS	0.002
	CgSNP180	379	2	0.364	0.375	0.480	0.115	0.182	*	0.122
	CgSNP150	381	2	0.363	0.496	0.476	0.113	0.181	NS	-0.021
	CgSNP91	367	2	0.362	0.695	0.476	0.113	0.181	NS	-0.187
	CgSNP225	383	2	0.361	0.514	0.474	0.112	0.181	NS	-0.041
	CgSNP65	382	2	0.356	0.382	0.464	0.107	0.178	*	0.095
	CgSNP126	368	2	0.356	0.554	0.464	0.107	0.178	NS	-0.089
	CgSNP52	383	2	0.355	0.334	0.462	0.106	0.177	*	0.059
	CgSNP28	383	2	0.352	0.418	0.457	0.104	0.176	NS	0.044
	CgSNP27	371	2	0.345	0.305	0.443	0.098	0.172	**	0.184
	CgSNP203	384	2	0.342	0.208	0.438	0.096	0.171	*	0.054
	CgSNP171	383	2	0.334	0.405	0.424	0.090	0.167	NS	0.012
Group 3	CgSNP164	384	2	0.332	0.396	0.420	0.088	0.166	NS	0.009
_	CgSNP34	382	2	0.327	0.120	0.413	0.085	0.164	*	0.047
	CgSNP206	381	2	0.322	0.346	0.404	0.082	0.161	NS	0.006
	CgSNP147	379	2	0.321	0.224	0.402	0.081	0.161	*	0.083
	CgSNP167	382	2	0.321	0.254	0.403	0.081	0.161	NS	0.012
	CgSNP155	379	2	0.320	0.050	0.400	0.080	0.160	NS	0.005
	CgSNP222	384	2	0.319	0.352	0.399	0.079	0.160	NS	0.002
	CgSNP187	376	2	0.317	0.250	0.396	0.078	0.159	*	0.125
	CgSNP209	377	2	0.312	0.475	0.387	0.075	0.156	NS	-0.103
	CgSNP197	380	2	0.308	0.132	0.381	0.072	0.154	*	0.085
	CgSNP148	370	2	0.307	0.232	0.378	0.071	0.154	*	0.037
	CgSNP07	380	2	0.300	0.089	0.367	0.067	0.150	NS	0.007
	CgSNP23	382	2	0.297	0.435	0.363	0.066	0.149	NS	-0.089
Group 4	CgSNP223	377	2	0.295	0.279	0.360	0.065	0.147	NS	-0.126
•	CgSNP220	382	2	0.294	0.421	0.359	0.064	0.147	NS	-0.080
	CgSNP111	373	2	0.283	0.324	0.342	0.058	0.142	NS	-0.025
	CgSNP183	380	2	0.283	0.400	0.342	0.058	0.142	NS	-0.079
	CgSNP232	383	2	0.283	0.420	0.341	0.058	0.141	NS	-0.104
	CgSNP162	382	2	0.274	0.199	0.329	0.054	0.137	*	0.045
	CgSNP194	383	2	0.271	0.405	0.323	0.052	0.135	NS	-0.111
	CgSNP36	353	2	0.266	0.348	0.317	0.050	0.133	NS	-0.048
	CgSNP41	379	2	0.264	0.361	0.313	0.049	0.132	NS	-0.072
	CgSNP35	383	2	0.249	0.355	0.292	0.043	0.125	NS	-0.095
	CgSNP149	375	2	0.227	0.277	0.262	0.034	0.114	NS	-0.029
	CgSNP176	380	2	0.202	0.216	0.229	0.026	0.101	NS	0.028

All SNPs displayed in decreasing order of PIC

N number of individuals, Na number of alleles, PIC polymorphic information content, H observed heterozygosity, He expected heterozygosity, EXCL1 and EXCL2 exclusion probabilities when no parent is known and when one parent is known, respectively, HWE significance of Hardy–Weinberg equilibrium test with Bonferroni correction, NS non-significant deviation from Hardy–Weinberg equilibrium, * and **, significant deviation from Hardy–Weinberg equilibrium, * (null) frequency of null allele calculated by Micro-checker 2.2.3



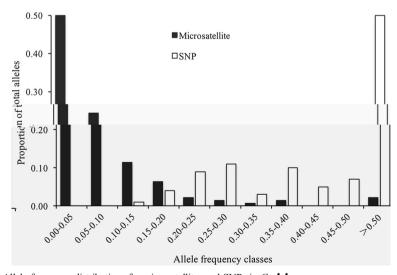
DNA dye, and melting instrumentation are required (Zhong et al. 2013). In this study, low genetic diversity of SNPs was revealed as the polymorphism information content (*PIC*) per locus ranged from 0.202 to 0.375 with an average of 0.328. In contrast, high genetic diversity in *C*. ** was presented using microsatellites with an average *PIC* value of 0.784. This discrepancy of *PIC* may be primarily due to the traits of markers. SNPs are bi-allelic while microsatellites are multiple alleles, thus SNPs shows lower information content than microsatellites (Aitken et al. 2004). Because of lower *PIC*, more SNPs are required than microsatellites to obtain the same power of exclusion.

All SNPs examined here were found to have an average *He* value of 0.417, which was significantly lower than the comparable value for the microsatellites (0.803). Vignal et al. (2002) predicted that SNP markers are mainly bi-allelic, such that a maximum expected heterozygosity value of 0.50 can be expected for a given SNP locus.

We found that most of the SNP alleles are at an intermediate or high frequency (Fig. 2). The differences in the frequency spectrum between microsatellites and SNPs may lead to the difference in the information content of the two types of marker (Kong et al. 2014).

The presence of null alleles was detected for pedigree analysis using microsatellite markers, which was in accordance with previous results in C. 2a (McGoldrick et al. 2000; Hedgecock et al. 2004; Li et al. 2009). This phenomenon could be attributed to indels or mutations on one or both of the primer-binding sites (Dakin and Avise 2004; Lemer et al. 2011; Mcinerney et al. 2011). High null allele frequency was also detected at four SNPs (CgSNP188, CgSNP180, CgSNP27, and CgSNP187). This uncommon problem was reported in Mediterranean mussel (M a a c c a) (Pino-Querido et al. 2015) as well. The same explanations presented above for high allele frequency at microsatellites may also be responsible for this observation at SNPs (Pino-Querido et al. 2015). In this study, some loci significantly deviated from Hardy–Weinberg (HW) proportions. These deviations resulted from difference between expected and observed heterozygosity. Null alleles are likely causes for heterozygote deficiency in HWE tests.

The level of unique assignment primarily depends on the assignment power of the marker set used, which is decided by the exclusion probabilities of the markers. Microsatellites were



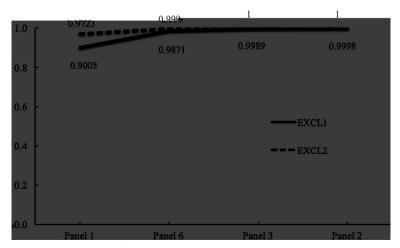
F g. 2 Allele frequency distributions for microsatellites and SNPs in C.



Tab e 4 Cumulative assignment success rates of simulated and real genotype data at the 95% confidence level. Each marker set was added in decreasing order of average polymorphic information content (*PIC*)

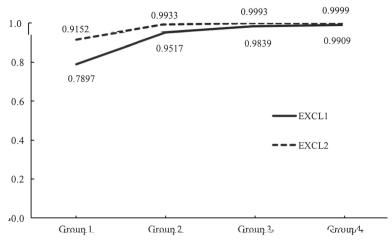
Type of markers	Cumulative assignment success (%)					
	Simulation data	Real data				
Microsatellite multiplex panels						
Panel 1	0	0				
Panel $(1 + 6)$	96	13				
Panel $(1 + 6 + 3)$	100	100				
Panel $(1 + 6 + 3 + 2)$	100	100				
SNP groups						
Group 1	0	0				
Group $(1+2)$	34	1				
Group $(1 + 2 + 3)$	100	98				
Group $(1 + 2 + 3 + 4)$	100	100				

individually more powerful to exclude a false parent than SNPs since EXCL1 and EXCL2 for each microsatellite were higher. The combined exclusion probabilities for both EXCL1 and EXCL2 as well as P_u were calculated by adding marker sets from highest average PIC value to lowest sequentially. The combined exclusion power for EXCL2 reached 1 when 3 microsatellites multiplex PCRs or more than 50 SNPs were used. Meanwhile, $P_{\rm u}$ was 0.9922 when three microsatellites multiplex panels were applied. With 50 SNPs, the same exclusion power could also be reached. In our study, the cost of microsatellites multiplex PCR analysis was estimated to be \$0.18 per sample per locus, while the cost of SNP genotyping was \$0.30. With increasing loci required to obtain a higher exclusion power, the genotyping costs would also increase. In practice, optimal investment in parentage assignment is a balance between the reduction of investment and operational costs needed for the separate family rearing and the cost of genotyping (Vandeputte and Haffray 2014). An increase in the number of selected loci resulted in an initial rapid rise in the exclusion probability value followed by slower relative growth until the maximum (100%) (Yu et al. 2015). To reduce the cost of pedigree analysis, we should explore the minimum and optimal criteria for microsatellites and SNPs required to attain a prerequisite statistical power in estimating pedigrees.



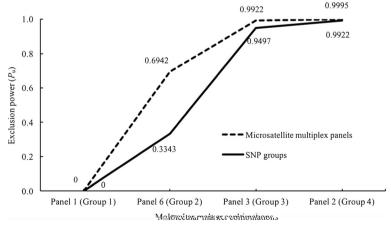
F g. 3 Combined exclusion probabilities of the four microsatellite multiplex panels analyzed for EXCL1 and EXCL2. Each panel was added in decreasing order of average polymorphic information content (*PIC*)





F g. 4 Combined exclusion probabilities of the four SNP groups analyzed for EXCL1 and EXCL2. Each group was added in decreasing order of average polymorphic information content (*PIC*)

The combined exclusion probabilities of 38 SNPs for EXCL2 came to 0.9993, which is nearly equal to the index of two microsatellites multiplex PCR panels. That was to say generally about six SNPs were needed to obtain an equivalent exclusion power for pedigree analysis with a microsatellite in *C.* *** a. Similarly, Glaubitz et al. (2003) estimated about six SNPs gave the same assignment power as 1 microsatellite. Simulations showed that about 16 bi-allelic SNPs would provide the same assignment power as three 10 allelic microsatellites (Wang 2006). In an inbred Augus herd, generally 2–3 SNPs per microsatellite were needed to obtain an equivalent exclusion power value (Fernández et al. 2013). Sellars et al. (2014) found the custom-made SNP panels were significantly cheaper to run while also providing faster turnaround time on genotyping results than multiplexed microsatellite panel. In sockeye salmon, Hauser et al. (2011) showed generally higher and more accurate parentage assignment success with 80 SNPs than 11 microsatellites. However, it is difficult to decide which type of marker is most cost-



F g. 5 Exclusion power (P_u) of four microsatellite multiplex panels and four SNP groups. Each marker combination was added in decreasing order of average polymorphic information content (PIC)

effective. Because SNPs remain more expensive due to the number required, but technology is rapidly evolving for SNPs and not for microsatellites for the moment (Vandeputte and Haffray 2014). The bright future of DNA markers applied for pedigree analysis will depend on the development of techniques. In fact, a few recent microsatellite studies relied on very large multiplexes (Guichoux et al. 2011). In *C.* *** , the largest multiplex panel simply contained five loci (Miller et al. 2012). Therefore, it is necessary to further develop multiplex PCRs to coamplify at least nine microsatellites in one reaction in *C.* *** . Simultaneous PCR amplification of several PCR products has been achieved in the case of SNPs (Gabriel et al. 2009; Vera et al. 2010; Pino-Querido et al. 2015). Helyar et al. (2011) also concluded multiple SNPs would be desirable to create minimum panels with maximum power for individual assignment. Therefore, development of SNP multiplex PCRs is worthy of consideration in *C.* *** to make the pedigree analysis more cost-effective.

Ac edge e • This study was supported by the grants from the National Natural Science Foundation of China (31372524), Shandong Seed Project, Shandong Province (2014GHY115002, 2016ZDJS06A06), and Qingdao National Laboratory for Marine Science and Technology (2015ASKJ02).

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