Estimating Breeding Values With Molecular Relatedness and Reconstructed Pedigrees in Natural Mating Populations of Common Sole, *Solea Solea*

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ABSTRACT

Captive populations where natural mating in groups is used to obtain offspring typically yield unbalanced population structures with highly skewed parental contributions and unknown pedigrees. Consequently, for genetic parameter estimation, relationships need to be reconstructed or estimated using DNA marker data. With missing parents and natural mating groups, commonly used pedigree reconstruction methods are not accurate and lead to loss of data. Relatedness estimators, however, infer relationships between all animals sampled. In this study, we compared a pedigree relatedness method and a relatedness estimator ("molecular relatedness") method using accuracy of estimated breeding values. A commercial data set of common sole, Solea solea, with 51 parents and 1953 offspring ("full data set") was used. Due to missing parents, for 1338 offspring, a pedigree could be reconstructed with 10 microsatellite markers ("reduced data set"). Cross-validation of both methods using the reduced data set showed an accuracy of estimated breeding values of 0.54 with pedigree reconstruction and 0.55 with molecular relatedness. Accuracy of estimated breeding values increased to 0.60 when applying molecular relatedness to the full data set. Our results indicate that pedigree reconstruction and molecular relatedness predict breeding values equally well in a population with skewed contributions to families. This is probably due to the presence of few large full-sib families. However, unlike methods with pedigree reconstruction, molecular relatedness methods ensure availability of all genotyped selection candidates, which results in higher accuracy of breeding value estimation.

^O estimate genetic parameters, additive genetic I relationships between individuals are inferred from known pedigrees (FALCONER and MACKAY 1996; LYNCH and WALSH 1997). However, in natural populations (RITLAND 2000; THOMAS et al. 2002) and in captive species where natural mating in groups is used to obtain offspring (BROWN et al. 2005; FESSEHAYE et al. 2006; BLONK et al. 2009) pedigrees are reconstructed. In these populations there is no control on mating structure, and typically unbalanced population structures with highly skewed parental contributions are obtained (BEKKEVOLD et al. 2002; BROWN et al. 2005; FESSEHAYE et al. 2006; BLONK et al. 2009). To reconstruct pedigrees, parental allocation methods are often used (MARSHALL et al. 1998; AVISE et al. 2002; DUCHESNE et al. 2002). These methods require that all parents be known. For situations where parental information is not available, numerous DNA-markerbased methods for estimating molecular relatedness have been developed (LYNCH 1988; QUELLER and GOODNIGHT 1989; RITLAND 2000; TORO et al. 2002).

These relatedness estimators determine relationship values between individuals on a continuous scale. Evaluation of relatedness estimators within real and simulated data in both plants and animals (*e.g.*, see VAN DE CASTEELE *et al.* 2001; MILLIGAN 2003; OLIEHOEK *et al.* 2006; RODRÍGUEZ-RAMILO *et al.* 2007; BINK *et al.* 2008) has generally focused on bias and sampling error of estimated genetic variances or relatedness values. Relatively little attention has been paid to their efficiency for estimation of breeding values.

Two types of relatedness estimators are currently available: method-of-moments estimators and maximum-likelihood estimators. Method-of-moments estimators (e.g., QUELLER and GOODNIGHT 1989; LI et al. 1993; RITLAND 1996; LYNCH and RITLAND 1999; TORO et al. 2002) determine relationships while calculating sharing of alleles between pairs in different ways. A variant of method-of-moments estimators is the transformation of continuous relatedness values to categorical genealogical relationships using "explicit pedigree reconstruction" (FERNÁNDEZ and TORO 2006) or thresholds (RODRÍGUEZ-RAMILO et al. 2007). However, correlations of transformed coancestries with known genealogical coancestries are low (RODRÍGUEZ-RAMILO et al. 2007). Several studies have compared different method-of-moments estimators but none revealed one

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single best estimator (VAN DE CASTEELE *et al.* 2001; OLIEHOEK *et al.* 2006; RODRÍGUEZ-RAMILO *et al.* 2007; BINK *et al.* 2008).

Maximum-likelihood (ML) approaches classify animals into a limited number of relationship classes (MOUSSEAU et al. 1998; THOMAS et al. 2002; WANG 2004; HERBINGER et al. 2006; ANDERSON and WEIR 2007). For each pair a likelihood to fall into a possible relatedness class (e.g., full sib vs. unrelated) is calculated given its genotype and phenotype. ML techniques combined with a Markov chain Monte Carlo approach reconstruct groups with specific relationships jointly and are therefore more efficient than other ML approaches. To minimize standard errors, all discussed ML methods require balanced population structures, large sibling groups, and a large variance of relatedness (THOMAS et al. 2002; WANG 2004; ANDERSON and WEIR 2007). Therefore, these methods may not be suitable for natural mating systems.

Unlike parental allocation methods, a benefit from relatedness estimators is that essentially all selection candidates are maintained for breeding value estimation, even with missing parents. The question is, however, whether such relatedness estimators also give accurate breeding values to perform selection.

In this study, we test suitability of a relatedness estimator to obtain breeding values in a population of common sole, Solea solea (n = 1953) obtained by natural mating. First, we estimate breeding values using pedigree relatedness of animals for which a pedigree could be reconstructed (using parental allocation). This data set (n = 1338) is further referred to as "reduced data set." We compare results with estimated breeding values using a simple but robust method-of-moments relatedness estimator: "molecular relatedness" (TORO et al. 2002, 2003). Next, we estimate breeding values using molecular relatedness in the full data set (n = 1953). Results show that accuracies of estimated breeding values obtained with molecular relatedness and pedigree relatedness are comparable. Accuracy increases when breeding values are estimated with molecular relatedness in the full data set. This implies that a molecular relatedness estimator can be used to estimate breeding values in captive natural mating populations.

MATERIALS AND METHODS

Data set: Two thousand animals (age 3 years old) from a commercial production population were sampled randomly from eight tanks. All animals have been produced by natural spawning and families were mixed over tanks. The parents (n = 51) had been collected from the Dutch North-Western coastal area and were sampled at the same time as the offspring. A blood sample for genotypic analysis was taken from all offspring and available parents. However, no samples were available from six potential parents due to mortality before sampling. On all offspring, body weight, gender, tank location, and measuring team were recorded. All animals

were genotyped with 10 microsatellite markers: *AF173855*, *AF173854*, *AF173852*, *AF173849*, *AY950593*, *AY950592* (IYENGAR *et al.* 2000), *AY950591*, *AY950589*, *AY950588*, *AY950587* (GAROIA *et al.* 2006). See BLONK *et al.* (2009) for details on DNA isolation and PCR amplification protocols.

Reconstructed pedigree relatedness: From the offspring data set, only individuals with more than five successfully amplified markers were used for analysis. Pedigree reconstruction was performed with PAPA 2.0 software (DUCHESNE *et al.* 2002). A uniform error of 0.02 on all markers was used. To set up a reliable pedigree, allocation results were checked as follows.

Offspring that were not allocated to any parent were removed from the data set. Also, animals that were allocated to more than two parents due to equal likelihood of breeding of these parents (termed as "ambiguous" in PAPA) were removed from the data set. Further, allocated parent– offspring pairs having the single highest likelihood of breeding among all other possible pairs were checked for consistency of Mendelian allele inheritance. Allocations were considered correct only when more than five markers were allocated consistently with Mendelian inheritance. Checking of allocation results was performed using R software (R DEVELOPMENT CORE TEAM 2008). The remaining data set was further referred to as a "reduced data set."

Molecular relatedness: Genetic relationships between animals were also estimated using a relatedness estimator. To estimate molecular relatedness, we used the method described by TORO *et al.* (2002, 2003). Here, coancestry (*f*) is calculated from similarity (*S*) of alleles between two individuals *x* and *y*. Here $S_{xy,l} = \frac{1}{4}[I_{ac} + I_{ad} + I_{bc} + I_{bd}]$, where at locus *l*, *a* and *b* are alleles of individual *x* and *c* and *d* are alleles of individual *y* (LI *et al.* 1993). When at I_{ac} allele *a* is identical to allele *c*, I_{ac} equals one, and zero otherwise, etc. According to LYNCH (1988) molecular coancestry is then calculated as

$$f_{xy} = \frac{1}{L} \sum_{l=1}^{L} \frac{S_{xy,l} - s_l}{1 - s_l},$$
(1)

where *L* is the number of markers (markers) and s_l is mean similarity (sum of squared allele frequencies, $\sum p^2$) at locus *l* in the base population. Estimated relatedness *r* between two animals is calculated as r = 2f. When ignoring alleles that are alike in state (AIS), molecular relatedness between two individuals is calculated from coancestry as follows:

$$r_{xy} = \frac{2}{L} \sum_{l=1}^{L} S_{xy,l}.$$
 (2)

Consequently, values of molecular relatedness are continuous and may range between 0 (no alleles are similar) and 2 (all alleles are similar).

Genetic analysis: To estimate heritabilities and breeding values, genetic analysis was performed with relatedness inferred from either pedigree reconstruction or molecular relatedness. The restricted maximum-likelihood method was used with ASReml software (GILMOUR *et al.* 2006) for analysis of the following genetic univariate linear model,

$$BW_{ijkl} = \mu + T_i + Ta_j + G_k + u_l + \epsilon_{ijkl}, \qquad (3)$$

where BW is the response variable for body weight, μ is the mean, T_{i} is the fixed effect of team *i*, Ta_j is the fixed effect of tank *j*, G_k is the fixed effect of gender *k*, u_l is the random animal effect of animal *l*, and ε is the random error term.

For estimation of breeding values from the pedigree reconstruction method, the relationship matrix was calculated within ASReml software using the reconstructed pedigree. The relationship matrix from the molecular relatedness method, however, was prepared separately and offered to ASReml as a generalized inversed matrix ("GIV-file"). To obtain convergence in ASReml, before inverting, the relationship matrix was made positive definite using the bending method. Bending and inverting was done using R software (R DEVELOPMENT CORE TEAM 2008). Heritabilities where then calculated using the estimated additive genetic variance and the residual variance as $\sigma_A^2/(\sigma_A^2 + \sigma_{E_1}^2)$.

Comparison of molecular relatedness and reconstructed pedigree relatedness: Pearson correlations were calculated between relatedness values as well as between breeding values obtained from molecular relatedness and reconstructed pedigree. To evaluate how well both methods fitted the model to the data, the accuracy in predicting the phenotype was calculated using cross-validation of each method. For crossvalidation, the data set was randomly divided into 100 equal sized subsets. Genetic analysis with the genetic model was then performed 100 times on all subsets jointly while successively omitting phenotype records for 1 subset at a time. In each analysis, for the subsets where observed phenotypes were omitted, phenotypes were predicted on the basis of available observed phenotypes in the other subsets.

Observed and predicted phenotypes from genetic analysis were corrected for fixed effects (referred to as "corrected phenotypes"). Accuracy of estimated breeding values ($r_{\rm HI}$) was calculated as

$$\frac{r_{\rm \hat{P}P}}{\sqrt{h^2}},\tag{4}$$

where $r_{\rm \tilde{PP}}$ is the Pearson correlation coefficient between predicted corrected phenotypes and observed corrected phenotypes observed from cross-validation and h^2 the heritability estimated using pedigree reconstruction. All calculations were performed using R (R DEVELOPMENT CORE TEAM 2008).

RESULTS

Data sets and relatedness: DNA isolation and marker analysis was successful, with only 30 samples giving no signal. Out of 2000 samples, finally 1953 animals could be genotyped with more than five markers. This is further referred to as the "full data set." In the full data set, on average 14.3 alleles per marker were found in the offspring. The parental data set, 13.4 alleles per marker were found.

After parental allocation, all allocated parent-offspring pairs were tested for consistency of Mendelian inheritance on each marker. The number of markers following correct Mendelian inheritance patterns was counted for each pair. The frequency distribution of allocated parent-offspring pairs over the number of consistent markers is shown in Figure 1. Forty-five percent of the pairs were allocated with all 10 markers. To construct the final pedigree, only parent-offspring pairs with more than 5 markers consistent with Mendelian inheritance patterns (66%) were taken into account. This resulted in a reconstructed pedigree for 1338 offspring, referred to as a reduced data set. In the reduced data set, 21 males and 17 females contributed to the offspring producing 59 full-sib families. Parental contribution was highly skewed with six parental pairs



FIGURE 1.—Frequency distribution of offspring allocated to two parents *vs.* number of markers that were allocated following Mendelian inheritance after pedigree reconstruction in cultured common sole (full data set, n = 1953).

producing 70% of the offspring. In the reduced data set, on average 13.7 alleles per marker were found in the offspring.

Parents were assumed to be unrelated. Consequently, relatedness values between offspring pairs after pedigree reconstruction were 0, 0.25, and 0.5. This corresponds to unrelated, half-, and full-sib pairs, respectively. The average relatedness in the reconstructed pedigree was 0.1483 ($\sigma_{PR} = 0.1865$). As expected, molecular relatedness values in the reduced data set were continuous and ranged between 0 and approximately 1.5. Average relatedness was 0.5588 ($\sigma_{MR} = 0.2100$). The continuous molecular relatedness values in the full data set ranged between 0 and 1.6. Average relatedness was 0.50 ($\sigma_{MR} = 0.2000$).

Genetic analysis: To obtain normally distributed residuals, the trait body weight was log-transformed. In the reduced data set, heritability for body weight was estimated at 0.23 (\pm 0.09) in a linear univariate model using relatedness inferred from pedigree reconstruction. The genetic variance was $0.56 \times 10^{-2} g^2$. With molecular relatedness in the reduced data set a heritability of 0.13 (\pm 0.04) was observed with a genetic variance of $0.31 \times 10^{-2} g^2$. In the full data set using molecular relatedness, heritability was estimated at 0.11 (\pm 0.03). Here, the observed genetic variance was $0.27 \times 10^{-2} g^2$.

Comparison of molecular relatedness and reconstructed pedigree relatedness: Per pedigree relationship class, the mean, standard deviation, and minimum and maximum values of molecular relatedness were calculated. Results are shown in Table 1. A large overlap of values obtained from molecular relatedness can be seen between relatedness classes of pedigree reconstruction. The Pearson correlation coefficient between molecular and reconstructed pedigree relatedness was $0.8 (P \leq 0.000)$. Estimated breeding values from pedigree reconstruction were positively correlated with breeding values obtained from molecular relatedness

Mean, standard deviation (SD), minimum (Min), and maximum (Max) of molecular relatedness (MR) between pairs of offspring per class of reconstructed pedigree relatedness (PR) in one generation of cultured common sole (reduced

data set, n = 1338)

PR class	MR mean	SD	Min	Max
Unrelated 0	0.4337	0.1254	0.0000	1.2500
Half sib 0.25	0.6354	0.1410	0.1000	1.4286
Full sib 0.5	0.8737	1.548	0.1875	1.5000
Self 1	1.2385	0.1216	1.0000	1.6000

(Figure 2). The estimated Pearson correlation coefficient was 0.77 ($P \le 0.000$).

Cross-validation with 100 equal sized randomly chosen subsets resulted in similar correlations between predicted and observed corrected phenotypes in both methods. In Figure 3 relationships between predicted and observed phenotypes in both methods are shown for the reduced data set. Note that the stratified shapes of Figure 3 can be explained by presence of large sib groups in the inferred pedigree. With pedigree reconstruction, predicted and observed phenotypes correlated with 0.2577 after correction for fixed effects. Accuracy of estimated breeding values with pedigree reconstruction was 0.54. Molecular relatedness resulted in a correlation of 0.2619 ($P \ll 0.000$) and an accuracy of 0.55. In the full data set, the correlation between predicted and observed phenotypes was 0.2936 ($P \ll$ 0.000) after correction for fixed effects. Consequently, the accuracy in the full data set was 0.60.

DISCUSSION

In this study we tested the suitability of molecular relatedness to estimate breeding values in a commercial population of common sole, *Solea solea*, obtained by natural spawning of parents. Progeny and parents were



FIGURE 2.—Relationship between breeding values estimated from molecular relatedness and reconstructed pedigree relatedness of offspring in cultured common sole (reduced data set, n = 1338).

genotyped with 10 microsatellite markers. Correlations of estimated relatedness and of breeding values between both methods were high. However, although relations were positive and significant, estimated heritabilities were different between methods.

In the reduced data set, accuracy of the univariate linear model was 0.54 with pedigree reconstruction and 0.55 with molecular relatedness (Figure 3). This result implies that both methods fitted the model to the data with nearly equal precision. In the full data set, accuracy with molecular relatedness increased to 0.60. These results show that in cases with missing pedigree data, molecular relatedness can be used to obtain breeding values. There was a highly skewed contribution of fullsib families to the offspring population. Therefore, the observed accuracies should be put in perspective of the maximum accuracy attainable with full-sib information (0.71). This implies that the observed levels of accuracy are high.

Prior to drawing conclusions on the optimal indicator of relationships between animals, it is useful to determine if the used forms of relationships add



FIGURE 3.—Relationship between predicted and observed corrected phenotypes in a linear model with reconstructed pedigree relatedness (A) and molecular relatedness (B) in one generation of cultured common sole (reduced data set, n = 1338).

information to breeding value estimation at all. To this purpose, we calculated breeding values using the model (Equation 3) without any relatedness information. Estimated breeding values from this model and from the model with relationships inferred from reconstructed pedigree correlated with 0.66 ($P \leq 0.000$, results not shown). This suggests that inclusion of relationships between animals alters estimation of breeding values.

Relatedness estimator: Many authors compared relatedness estimators under different population structures, numbers of markers, and alleles (*e.g.*, VAN DE CASTEELE *et al.* 2001; OLIEHOEK *et al.* 2006; BINK *et al.* 2008). ML estimators (MOUSSEAU *et al.* 1998; THOMAS *et al.* 2002; WANG 2004; HERBINGER *et al.* 2006) or pedigree reconstruction methods that do not need parental information (BERGER-WOLF *et al.* 2007; ASHLEY *et al.* 2009) were less suitable for our situation. Assumptions made by these methods, as balanced mating structures and large sibling groups, would be violated in our data set. In the data set used, skewed parental contributions were observed (data not shown). This is typical for natural spawning species in, *e.g.*, aquaculture (BROWN *et al.* 2005; BLONK *et al.* 2009).

From the literature it remains unclear which methodof-moments relatedness estimator performs best; differences are small. In this study we used molecular coancestry (TORO *et al.* 2002; TORO *et al.* 2003): among all estimators compared, molecular coancestry turned out to be relatively simple but robust.

The importance of the number of markers for the power of a relatedness estimator has been emphasized by several authors. OLIEHOEK et al. (2006) and BINK et al. (2008) showed that reasonable correlations (>0.7)between estimated and pedigree relatedness were found only in simulated data when using at least 50 markers and four to five alleles per marker. In contrast, in this study a correlation of 0.8 was found between estimated and pedigree relatedness using 10 markers with 13.7 alleles per marker on average. The difference in results might also be caused by the characteristics of the population under study. Relatedness estimators generally are used for natural populations in which variation of relationship structures is low. In these cases discriminative power of relatedness estimators is low with few markers. For example, THOMAS et al. (2000) and WILSON and MCDONALD (2003) showed that 12 to 15 markers were not enough to produce reliable heritability estimates with a relatedness estimator. The tested populations (Soay sheep, Ovis aries, and rainbow trout, Oncorhynchus mykiss) showed low variation of relationship structures. Larger variation of relationship structures, i.e., larger families but also larger sample sizes, decreases bias and sampling errors of estimated relatedness (THOMAS et al. 2000; THOMAS et al. 2002; RODRIGUEZ-RAMILO et al. 2007). In aquaculture populations with natural spawning in groups, generally

large variation of relationship structures is found (GJEDREM 2000; VANDEPUTTE 2005; SAILLANT *et al.* 2006). As contribution of families to the offspring population in this study was highly skewed, this may explain the success with relatively few markers.

Heritability: In this study heritability for body weight in common sole was estimated at 0.23 (± 0.09) when using pedigree relatedness. The estimated heritability is in line with values found for bodyweight in other aquatic species (GJEDREM 2000). Heritability estimated from molecular relatedness, however, was much lower: 0.11 (± 0.02) -0.13 (± 0.04) . This suggests underestimation of genetic variance, which is likely to be caused by erroneous ascribing of relationships between offspring due to too few markers. THOMAS et al. (2002) showed that, among other factors, use of fewer markers increases variance and bias of sampling error variance of estimated relatedness. With higher variance of relatedness, genetic variance and heritability decrease as $\hat{\sigma}^2_{A} r =$ $cov(P_i, P_i)$ (FALCONER and MACKAY 1996). This is supported by a study on genomic selection in which fewer markers led to underestimation of heritability estimates in an Angus cattle population (HAYES and GODDARD 2008). A similar effect was demonstrated by SHIKANO (2008) in Japanese flounder, Paralichthys olivaceus. It is expected that heritability will be estimated more accurately with a higher number of markers and consequently will approach the value from pedigree reconstruction.

In our particular case of skewed contribution of parents, the observed heritabilities from molecular relatedness will have little effect on estimated breeding values and accuracies. This is because breeding values were based on family information (*i.e.*, family means) rather than on heritabilities.

Applications for breeding programs: Our study illustrates that pedigree reconstruction using genotyping data is an inefficient method when parental genotypes are missing. Difficulty with allocation is reflected in Figure 1. Loss of selection candidates due to ambiguous parental allocation was 33%. Comparable results (10–30%) were found by several authors in other species (HERLIN *et al.* 2007, 2008; PIERCE *et al.* 2008). Loss of selection candidates has major effects on costs of selection procedures in breeding programs and may lead to lower realized selection responses and increased rates of inbreeding.

As parental allocation programs often allow uncertainties to the performed allocations (DUCHESNE *et al.* 2002), mistakes in reconstructed pedigrees may occur. This increases the risk that related animals are considered as unrelated. Selection of animals using optimal contribution theory (MEUWISSEN 1997) may therefore unintentionally increase rates of inbreeding.

This study shows that the molecular relatedness estimator circumvents problems with pedigree reconstruction in an aquaculture population with natural spawning of groups of parents. Moreover, due to preservation of all selection candidates, a higher accuracy of breeding value estimation is achieved. The increasing amount of genetic information (*e.g.*, number of SNPs and microsatellites) made available due to current developments in genomics (HAYES *et al.* 2009) will further enhance accurate estimation of genetic parameters.

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