Random regression models for genetic evaluation of performance of the Hungarian show-jumping horse population

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SUMMARY

The aim of the study was to estimate genetic parameters for show-jumping competition performance using random regression model. Show-jumping competition results collected between 1996 and 2009 were analyzed. The database contained 272 951 starts of 8020 horses. Identity number and gender of the horse, rider, competition date, the level of the competition and placing were recorded in the database. Competition levels were categorized into five groups. Weighted – competition level used – square root transformed placing was used to measure performance of horses. The random regression model included fixed effects for gender, year and place of competition, and random effects for rider, animal and permanent environment.

Later performance of show-jumping horses measured with weighted square root ranks is less influenced by rider and permanent environmental effects than performance at the beginning of a horse’s sporting career. Heritability increased continuously from 6.3 years of age (2296 age in days), values were in the range of 0.07 and 0.37. Higher heritability was found in later ages. Weak genetic and phenotypic correlation was found between the early 4–5–6 years of age and older (7, 8, 8+) age classes. From 8.5 years of age (3132 days old) there were strong genetic and phenotypic correlations between neighboring age groups. For the same age classes moderate and strong genetic and phenotypic correlation was found. Genetic correlation between 13.5 years of age and older horses was very strong.

Keywords: random regression, sporthorse, performance measurement

INTRODUCTION

Sporthorse breeder’s interest is focused on changes in performance within individual horses over their career. Changes can be observed not only in means, but there can be heterogeneity in variances and covariances at different ages. The way how to investigate changes in variances and covariances is by defining appropriate covariance functions (Kirkpatrick et al., 1990). Covariance functions are continuous functions that should make possible to estimate covariance within animals between measures at any two age $t_i$ and $t_j$, and variances between animals at any age $t_i$.

To estimate covariance functions an alternative is using random coefficient regression model. Thus the regression coefficients are assumed to be random variables and there is a covariance structure among regression coefficients. These can be used to estimate covariance functions, i.e. variances and covariances between values for different ages.
with time, because of improved modelling of variances and genetic parameters (Meyer, 2004). RRM provides higher volume of information than the repeatability animal model (Posta et al., 2010). The use of random regression model is highly recommended as a very useful tool, because it allows us to estimate the animal's breeding value through all the stages of its trajectory in racing competitions.

In the last decade some studies used this methodology in horses (Kaps et al., 2004; Bugislaus et al., 2006; Menendez-Buxadera and Mota, 2008; Posta et al., 2010; Gómez, 2011).

Kaps et al. (2004) investigated changes in variances and covariances of withers height of Lipizzan horses from birth to 36 months of age. Bugislaus et al. (2006) analysed the trajectory of racing time per kilometre by age (6 classes: 2, 3, 4, 5, 6 and over 6 year olds) in German Trotter Horses; whereas Menendez-Buxadera and Mota (2008) analysed the complete race time (in seconds) by distance (7 classes: 1000–1600 m) in Brazilian Thoroughbreds. Posta et al. (2010) estimated genetic parameters for show-jumping performance using shifted Blom score and difference between fence height and fault point as performance measurement variable. Gómez et al. (2011) have analyzed the changes in the trajectory of accumulated earnings for Spanish Trotter Horses during their sporting life.

The aim of the study was to estimate genetic parameters for performance of Hungarian show jumping horses during their career using random regression model.

**MATERIAL AND METHODS**

Show-jumping competition results collected between 1996 and 2009 were analyzed. The data used in this study were obtained from the Hungarian Equestrian Federation. The final dataset contained in total 272,951 competition records on 80,200 individual horses after data screening, results were gathered from Hungary and other European countries. Identity number and gender of the horse, rider, year, level and location of the competition and placing were recorded in the dataset. Information about pedigree of horses were gathered and set up with help of the National Horse Breeder Information System. The pedigree file contained 40,142 animals four generation back.

Competitions were categorized into five groups based on their difficulty level. For the evaluation of show-jumping performance weighted square root ranks was used. Analysis before showed that the best goodness-of-fit value were in case of weighted square root (Rudiné et al., 2013). Sporthorses were divided into 6 age groups, 4, 5, 6, 7, 8 and more than 8 years old.

To estimate variances and covariances of regression coefficients Legendre polynomials were used (Kirkpatrick et al., 1990). The random regression coefficients for additive genetic and permanent environmental effects were fitted with first to fifth (n=1, 2, 3, 4, 5) orders of Legendre polynomial for the trait. The general form of a Legendre polynomial is:

\[ \phi_n(t) = \frac{1}{2^n} \sqrt{\frac{2n+1}{2}} \sum_{k=0}^{\left\lfloor \frac{n}{2} \right\rfloor} (-1)^k \binom{n}{k} \binom{2n-2k}{n} q_t^{n-2k} \]

where \( q_t = -1 + 2 \frac{t-t_{\min}}{t_{\max} - t_{\min}} \), \( n \) is the degree of the Legendre polynomial, \( k=0,...,n \), and \( q_t \) is the standardized time converted to a number between -1 and 1.

To estimate the trajectory of the variance components the following random regression model was used applying the VCE-6 program developed by Kovac and Groeneveld (2003):

\[
Y_{jklmno} = \text{Gender}_j + \text{Year}_k + \text{Place}_l + \text{Rider}_m + \sum_{n=1}^{\infty} t^n + \sum_{n=1}^{\infty} \alpha_n \phi_n(q_t) + \sum_{n=1}^{\infty} \gamma_n \phi_n(q_t) + e_{jklmno}
\]

where:
- \( Y_{jklmno} \) is a vector of observations for the particular trait of a specific horse,
- \( \text{Gender}_j \) is fixed effect of gender (3 levels),
- \( \text{Year}_k \) is fixed effect of year of competition (14 levels),
- \( \text{Place}_l \) is fixed effect of place of competition (443 levels),
- \( \text{Rider}_m \) is random effects of the rider (3352 levels),
- \( t = \) age in days,
- \( \phi_n = \) the Legendre polynomial,
- \( \alpha_n = \) random regression coefficients of additive genetic effect,
- \( \gamma_n = \) random regression coefficients of permanent environmental effect,
- \( e_{jklmno} = \) random residual effect.

The regression coefficients are assumed to be random variables. Regressions were split additive genetic effect regression and permanent environment regression due to a horse. Gender of the horse, year, place of the competition were defined as fixed effect.
After examining the estimated variances and variance proportions for all models, the best fitting random regression model was the model with a polynomial of first order of fit (LP 1) for all random effects.

Estimates of the first eigenfunction for weighted square root ranks were positive from age 6.08 (2219 age in days) (Figure 2) indicating that selection on the sport performance traits could be successful from 2219 days old. In other words selection on this trait at any age from 6.08 should change show-jumping performance in the same direction. For weighted square root ranks the first eigenvalue explained 99.56% of the genetic variation, and the second eigenfunction was responsible for only 0.44% of the total genetic variation for the first ordered Legendre polynomial. Variances explained by eigenfunctions of second order Legendre polynomials were lower, so the first order LP 1 was the best fitting.

The heritability values were in the range of 0.07 and 0.37 (Figure 4), it increased continuously after 6.3 years of age (2296 age in days). Variance proportion of permanent environmental effect decreased continuously up to 8.3 years of age than it increased slightly over time. Proportion of residual variance increased up to 6 years of age and afterwards its effect was lower and lower over time. Proportion of variance for random effect of rider continuously decreased.

Heritabilities based on square root of placing were similar to those of Huizinga and van der Mei (1989) $h^2=0.10–0.20$; Koenen et al. (1995) $h^2=0.17$; Hassenstein et al. (1998) $h^2=0.07–0.11$; Luehrs Behnke et al. (2002) $h^2=0.11$; Jaitner and Reinhardt (2003) $h^2=0.10$; Sobczynska and Lukaszewicz (2004) $h^2=0.15$; and Viklund et al. (2011) $h^2=0.11$. Posta et al. (2010) estimated 0.037 and 0.385 heritabilities of shifted Blom score.

From 8.5 years of age (3132 days old) there were strong genetic and phenotypic correlations between neighboring age groups (Figure 5). For the same age classes moderate and strong genetic and phenotypic correlation was found. Weak genetic and phenotypic correlation was found between the early 4–5–6 years of age and older (7, 8, 8+) age classes. Genetic correlation between 13.5 years of age and older horses was very strong.
CONCLUSIONS

Performance reached at younger age (4–5–6 years old horses) do not reflect the later performance at all. Weak genetic and phenotypic correlation was found between the early 4–5–6 years of age and older (7, 8, 8+) age classes. From 6 years of age selection could be successful on weighted square root ranks. Later performance of show-jumping horses measured with weighted square root ranks is less influenced by rider and permanent environmental effects than performance at the beginning of a horse’s sporting career. Heritabilities increased continuously from 6.3 years of age (2296 age in days), values were in the range of 0.07 and 0.37. Higher heritabilities were found in later ages. From 8.5 years of age (3132 days old) there were strong genetic and phenotypic correlations between neighboring age groups. For the same age classes moderate and strong genetic and phenotypic correlation was found. Genetic correlation between 13.5 years of age and older horses was very strong.

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