

1 **Exploring the potential genetic heterogeneity in the incidence of hoof and leg**
2 **disorders in Austrian Fleckvieh and Braunvieh cattle**

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13

14 **Abstract**

15 **Background:** Genetic heterogeneity denotes the situation when different genetic architectures
16 underlying diverse populations result in the same phenotype. In this study, we explore the nature
17 of differences in the incidence of the number of hoof and leg disorders between Braunvieh and
18 Fleckvieh cattle in the context of genetic heterogeneity between the breeds.

19 **Results:** Despite potentially higher power of testing due to twice as large sample size, none of
20 the SNPs was significantly associated with the number of hoof and leg disorders in Fleckvieh,
21 while 16 SNPs were significant in Braunvieh. The most promising candidate genes in
22 Braunvieh are: CBLB on BTA01, which causes arthritis in rats; CAV2 on BTA04, which in
23 effects mouse skeletal muscles; PTHLH on BTA05, which causes disease phenotypes related
24 to the skeleton in humans, mice and zebrafish; SORCS2 on BTA06, which causes decreased
25 susceptibility to injury in the mouse. Some of the significant SNPs (BTA01, BTA04, BTA05,
26 BTA13, BTA16) reveal allelic heterogeneity – i.e. differences due to different allele frequencies
27 between Fleckvieh and Braunvieh. Some of the significant regions (BTA01, BTA05, BTA13,
28 BTA16) correlate to inter-breed differences in LD structure and may thus represent false-
29 positive heterogeneity. However, positions on BTA06 (SORCS2), BTA14 and BTA24 mark
30 Braunvieh-specific regions.

31 **Conclusions:** We hypothesise that the observed genetic heterogeneity of hoof and leg disorders
32 is a by-product of multigenerational differential selection of the breeds – towards dairy
33 production in the case of Braunvieh and towards beef production in the case of Fleckvieh. Based
34 on the current data set it is not possible to unequivocally confirm/exclude the hypothesis of
35 genetic heterogeneity in the susceptibility to leg disorders between Fleckvieh and Braunvieh
36 because only explore it through associations and not the causal mutations. Rationales against
37 genetic heterogeneity comprise a limited power of detection of true associations as well as
38 differences in the length of LD blocks and in linkage phase between breeds. On the other hand,
39 multigenerational differential selection of the breeds and no systematic differences in LD
40 structure between the breeds favour the heterogeneity hypothesis at some of the significant sites.

41

42 **Keywords:** Braunvieh, feet and leg disorders, Fleckvieh, genetic heterogeneity, GWAS,
43 linkage disequilibrium, principal components

44 **Background**

45 Genetic heterogeneity denotes the situation when different genetic architectures underlying
46 diverse populations result in the same phenotype. In human genetics, for decades, the concept
47 of genetic heterogeneity has been considered in genome-wide association studies (GWAS) [1].
48 One of the most well-known diseases characterised by high degree of genetic heterogeneity is
49 the human autism spectrum disorder [2]. Relatively recently the concept of genetic
50 heterogeneity has also been introduced into to the analysis of data from artificially selected
51 plant and livestock species by Béréños et al. [3], de los Campos et al. [4], and Lehermeier et al.
52 [5]. In such species, an important source of genetic heterogeneity may be due to a complex
53 population structure, which is typically composed of divergently selected breeds exhibiting
54 high variation in allele frequencies and linkage disequilibrium patterns [6].

55 Cattle hoof and leg disorders are relatively novel traits represented by a group of different
56 phenotypes varying from binary, directly assessed disease diagnoses like e.g. a sole ulcer to
57 composite traits scored on a categorical basis, e.g. a locomotion score. Due to their impact on
58 welfare, productivity and fertility [7], the traits rapidly gain importance in dairy cattle breeding
59 schemes. Technically, a common feature is a relatively poor definition of traits from this group
60 and lack of routine recording, resulting in a large number of relatively small data sets scattered
61 across various populations. Those features not only cause low power of detection of significant
62 gene (or SNP) – phenotype associations resulting in a low reproducibility of results [7], but also
63 imply a potential heterogeneity in the genetic determination of phenotypes due to differences
64 in selection schemes and thus underlying differences in linkage disequilibrium and allele
65 frequencies between populations [8].

66 In this study, we explore the nature of differences in the incidence of the number of hoof and
67 leg disorders between Austrian bred Braunvieh and Fleckvieh cattle in the context of genetic
68 heterogeneity between the breeds.

69 **Results**

70 **Heterogeneity in association signals**

71 Adapting the false discovery rate (FDR) threshold of 10%, despite potentially higher power of
72 testing due to twice as large sample size, none of the SNPs was significantly associated with

73 the number of hoof and leg disorders in Fleckvieh, while 16 SNPs were significant in Braunvieh
74 (Figure 1). One of the three significant SNPs from BTA01 is located 285,955 bp upstream of
75 CBLB gene, known to cause arthritis in rats. The same SNP is located within a region of a QTL
76 for hindquarter proportions. The two most significant SNPs were located on BTA04. Both were
77 intergenic, but their closest downstream gene encodes caveolin 2 protein, which in the mouse
78 is known to effect skeletal muscles. A SNP on BTA05 falls within four QTL regions responsible
79 for rump conformation traits. One of the most interesting significant annotation points at
80 another SNP on BTA05, which is located 76,362 bp downstream of PTHLH. In humans and
81 mice, this gene causes multiple disease phenotypes related to the skeleton. In zebrafish
82 mutations of this gene result in decreased bone mineralisation, in humans – to brachydactyly
83 and to numerous bone and calcium related disease phenotypes in the mouse, including:
84 decreased length of long bones, premature bone ossification, and increased osteocyte apoptosis.
85 Another interesting significant annotation points at the intron of SORCS2 gene on BTA06,
86 which was assigned to decreased susceptibility to injury phenotype in the mouse. The effect on
87 muscles, albeit in Zebrafish, was assigned to the protein encoded by PIP4K2A, which is located
88 close to the significant SNP on BTA13. The same SNP is also within a QTL region for rump
89 angle. On BTA16, a significant association points out at ENSBTAG00000009943 involved in
90 inflammatory response. Significant SNPs are summarised in Table 1 except a SNPs on BTA06,
91 which could not be placed on the current reference assembly (ARS-UCD1.2).

92 There was no correlation between P values observed for SNPs in FLV and BSW, which was
93 estimated to 0.00302 for all SNPs and -0.01494 (-0.08065) for SNPs with 100 smallest P values
94 in BSW (FLV). In addition, breed-specific SNP effect estimates also revealed a very low
95 correlation of 0.02363.

96 **Heterogeneity in genetic architecture**

97 For the chromosomes containing SNPs significant in Braunvieh, Machalanobis distances,
98 expressing differences between breeds in SNP genotype variability, and the S statistics,
99 expressing differences in the LD decay pattern, were visualised on Figure 2. For the
100 Machalanobis distance, all FDR values at the significant SNP locations are equal to one. Even
101 while considering whole chromosomes harbouring significant SNPs none of the distances was
102 significant, indicating that it was not possible to differentiate between breeds based on SNP
103 genotypes corresponding to the 50-SNP windows. A somewhat different picture emerged when
104 inter-breed differences in LD were considered. In some, but not all, of the regions, significant

105 SNPs correspond to windows for which a difference in LD structure was indicated by high
106 values of the *S* statistics - rs110488513 and rs41661497 on BTA01 as well as rs110792762 on
107 BTA13. Some other significant SNPs are located in windows adjacent to such windows -
108 rs110811919 on BTA1, as well as rs29024589, rs110843300, and rs41579631 on BTA16. For
109 eight significant SNPs (rs29024589 on BTA01, both SNPs on BTA04, both SNPs on BTA05,
110 both SNPs on BTA13, rs41579631 on BTA16) significant allelic heterogeneity was detected
111 (Table 1).

112 Summarising the obtained results, inter-breed differences at some of the 16 significant
113 positions can be explained by inter-breed differences in LD structure (BTA01, BTA05, BTA13
114 and BTA16) indicated by high values of the *S* statistics and/or by significant allelic
115 heterogeneity (BTA01, BTA04, BTA05, BTA13 and BTA16). Still polymorphisms on BTA06
116 (marking SORCS2), BTA14 (marking TMEM74 and EMC2) and BTA24 (marking CCDC178
117 and KLHL14) are good candidates for Braunvieh-specific associations.

118 **Discussion**

119 Although the number of published studies related to GWAS for hoof and leg disorders is very
120 limited, their common feature is the lack of overlap in significant results both, between and
121 even within the studies. Similarly, to our study, Wu et al. [9] applied the same GWAS model
122 to feet and leg disorders in three breeds and depending on breed identified different significant
123 regions between Danish Red and Danish Holstein, while no significance was observed in
124 Jersey. In addition, earlier van der Spek et al. [10] found a very low overlap in significance
125 while analysing a cow data set and a bull data set ascertained from the same population of
126 Holstein-Friesian cattle, with only three SNPs in bulls overlapped with 94 SNPs significant for
127 claw disorders in cows. Vargas et al. [11] found no overlap between significant regions defined
128 for binary and categorical feet and leg classification scores in Nelore breed.

129 Also, in our study we observed no overlap in significance between Braunvieh and Fleckvieh.
130 The potential basis of this phenomenon is either of a technical nature – type I/type II errors due
131 to limited sample size, or of a genetic nature - genetic heterogeneity in the susceptibility to leg
132 diseases between breeds. In humans, Coram et al. [12] reported a similar result regarding loci
133 determining concentration of lipids in blood, where many differences between populations were
134 due to allele frequencies at the candidate SNPs. The same study also points out at the presence
135 of population-specific significant loci, which, as in our study, can be explained by population-

136 specific selection pressure. Another postulated cause of heterogeneity, see e.g. An and
137 Claudianos [2], for their discussion on autism disorder), pointing out at different causal
138 mutations within the common metabolic pathways. Similarly, Wu et al. [9] in the context of
139 hoof and leg disorders in cattle hypothesised that the breed-specific significance hits represent
140 relatively novel mutations, which occurred after breed separation. The third cause of
141 heterogeneity are differences in genetic architecture between breeds, which are manifested by
142 genome-wide (Figure 3), but also by local differences in LD, which were detected in our study
143 within some of the regions harbouring SNPs significant in Braunvieh. Differences in LD
144 patterns were considered in the context of heterogeneity detected between human populations
145 [11].

146 **Conclusions**

147 Based on the current data set it is not possible to unequivocally confirm/exclude the hypothesis
148 of genetic heterogeneity in the susceptibility to leg disorders between Fleckvieh and Braunvieh.
149 The rationales against the hypothesis comprise: (i) limited power of detection of true
150 associations if the effect size is not large and therefore high rate of spurious associations among
151 detected SNPs, (ii) differences in the length of LD blocks, which imply differences in power of
152 detecting the associations, (iii) differences in linkage phase between breeds, which may hamper
153 the detection of causal sites in Fleckvieh or Braunvieh based on the available SNP panel. On
154 the other hand (i) multigenerational differential selection of the breeds – towards dairy
155 production in the case of Braunvieh and towards beef production in the case of Fleckvieh, (ii)
156 no significant allelic heterogeneity, and (iii) no systematic differences in LD structure between
157 the breeds stay in favour of the heterogeneity hypothesis at the significant sites on BTA06,
158 BTA14, and BTA24.

159 Unfortunately, the data set available for the analysis comprises only common SNPs selected
160 for a commercial microarray, so that we can explore only associations and not the causal
161 mutations, therefore a final verification of the above hypothesis would require a denser SNP
162 map from whole genome sequence.

163 **Methods**

164 **Dataset**

165 The analyzed data set was collected within the frame of the Efficient Cow project and comprised
166 scores of hoof and leg disorders in Austrian 985 Braunvieh and 1,999 Fleckvieh cows. In
167 particular, the analyzed phenotype comprised the total number of hoof disorders scored until
168 100th day of lactation. In both breeds, the number of disorders varied between none and five,
169 but the distributions differ with the fraction of diseased cows being higher in Fleckvieh (Figure
170 1). The cows were genotyped with the GeneSeek[®] Genomic ProfilerTM HD panel consisting
171 of 76,934 SNPs out of which 74,762 SNPs remained for further analysis after preprocessing
172 based on a minor allele frequency (<0.01) and a per-individual call rate (<99%).

173 **Genome-wide association study**

174 The genome-wide association study was performed separately for each breed by applying a
175 series of single-SNP mixed linear models implemented in the GCTA software [13] for
176 pseudophenotypes, expressed by cows' breeding values estimated by Suchocki et al. [14]. For
177 a single SNP the model is given by:

$$178 \quad \mathbf{u} = \mu + \mathbf{X}\mathbf{b} + \mathbf{Z}\mathbf{g} + \mathbf{e} ,$$

179 where, \mathbf{u} is a vector of breeding values, μ is a general mean, \mathbf{b} is the fixed additive effect of a
180 single SNP, \mathbf{X} is a corresponding design matrix coded as 0, 1 or 2 for a homozygous,
181 heterozygous and the other homozygous genotype respectively, $\mathbf{g} \sim N(0, \mathbf{G}\sigma_g^2)$ is a random
182 additive polygenic effect with the genomic covariance matrix between cows (\mathbf{G}) calculated
183 based on SNP genotypes, \mathbf{Z} is an incidence matrix for \mathbf{g} , $\mathbf{e} \sim N(0, \mathbf{I}\sigma_e^2)$ is a residual. The null
184 hypothesis of $\mathbf{b} = 0$ was tested using the Likelihood Ratio Test with the asymptotic large
185 sample χ_1^2 distribution (as implemented into the GCTA). The resulting nominal P-values were
186 transformed into False Discovery Rates [15] to account for multiple testing.

187 **Analysis of allelic heterogeneity**

188 For each, non-overlapping windows of 50 neighboring SNPs a genomic relationship matrices
189 between cows were calculated, which were then decomposed into principal components, using
190 the PCA subroutine implemented into GCTA [16]. Further on, for each of the windows,
191 differences between the breeds in a 10-dimensional space defined by the first ten eigenvectors
192 ($\boldsymbol{\varepsilon}_1, \boldsymbol{\varepsilon}_2, \dots, \boldsymbol{\varepsilon}_{10}$) were quantified using the Mahalanobis distance: $D_M = \sqrt{\mathbf{d}'\mathbf{V}^{-1}\mathbf{d}}$. With $\mathbf{d} =$
193 $[\bar{\varepsilon}_{1B} - \bar{\varepsilon}_{1F}, \bar{\varepsilon}_{2B} - \bar{\varepsilon}_{2F}, \dots, \bar{\varepsilon}_{10B} - \bar{\varepsilon}_{10F}]$ containing differences between averaged eigenvectors
194 for Braunvieh (subscript B) and Fleckvieh (subscript F) and \mathbf{V} representing a pooled covariance
195 matrix of $\boldsymbol{\varepsilon}_1$ and $\boldsymbol{\varepsilon}_2$. The Hotelling test: $T = \frac{n_B n_F}{n_B + n_F} \cdot \frac{n_B + n_F - 11}{10(n_B + 2n_F - 2)} \cdot \mathbf{d}'\mathbf{V}^{-1}\mathbf{d} \sim F_{10, n_B + n_F - 11}$ was

196 used to test the null hypothesis of no differences in positions between Braunvieh and Fleckvieh,
197 where n_x is the number of cows representing each breed [17].

198 The allelic heterogeneity between breeds was tested by calculating the ratio of minor allele
199 frequencies in Fleckvieh (MAF_F) and Braunvieh (MAF_B) at significant SNP positions. For
200 hypotheses testing, the large sample standard normal distribution of the natural logarithm of the
201 ratio was used: $\ln\left(\frac{MAF_F}{MAF_B}\right) \sim N(0,1)$.

202 **Analysis of local linkage disequilibrium patterns**

203 Differences in linkage disequilibrium (LD) patterns between breeds were assessed based on the
204 comparison of LD matrixes constructed for non-overlapping windows of 50 neighboring SNPs.
205 LD between pairs of linked SNPs was quantified using Beagle 4.1 [18], separately for each
206 breed, by the r^2 coefficient given by $\frac{(p_{11}p_{22}-p_{12}p_{21})^2}{p_{1.}(1-p_{1.})p_{.1}(1-p_{.1})}$, with p_{ij} corresponding to the frequency
207 of a two-SNP haplotype $ij \in \{11,12,21,22\}$, $p_{1.}$ and $p_{.1}$ representing the marginal SNP allele
208 frequency. Eigenvectors corresponding to the LD matrices were computed separately for each
209 breed, using Python scripts. Inter-breed differences in local linkage disequilibrium were then
210 quantified by:

$$211 \quad S = 2 \sum_{i=1}^{50} [(v_{iBB} - v_{iFB})^2 + (v_{iBF} - v_{iFF})^2] ,$$

212 where v_{ijk} corresponds the i -th element the 1st principal component vector calculated as the
213 product of the LD matrix of the j -th breed (subscript B for BSW or subscript F for FLV) and
214 the 1st eigenvector of the k -th breed. Following Garcia [19], S quantifies differences in the
215 variability of LD between two populations. Furthermore, the genome-wide pattern of linkage
216 disequilibrium (LD) decay with physical distance of pair-wise SNPs were binned into seven
217 types of intervals (0 to 25 kb, 25 to 50 kb, 50 to 100 kb).

218 **List of abbreviations**

219 BSW: Braunvieh, FDR: false discovery rate, FLV: Fleckvieh, GWAS: genome-wide
220 association study, LD: linkage disequilibrium, PCA: principal component analysis, SNP: single
221 nucleotide polymorphism

222 **List of tables**

223 Table 1. SNPs significant in BSW, based on the $FDR \leq 0.10$ threshold. SNP genomic
224 information corresponds to the ARS-UCD1.2 reference genome.

225 **List of figures**

226 Figure 1. Manhattan plots for GWAS for hoof and leg disorders in Braunvieh and Fleckvieh.
227 The horizontal line corresponds to FDR of 0.01.

228 Figure 2. Descriptive statistics of heterogeneity between Braunvieh and Fleckvieh on
229 chromosomes containing SNPs significant for Braunvieh in GWAS. Vertical bars mark the
230 positions of significant SNPs.

231 Figure 3. Genome-wide LD decay in Braunvieh and Fleckvieh.

232 **Availability of data and materials**

233 The data that support the findings of this study are available from the ZuchtData EDV-
234 Dienstleistungen GmbH but restrictions apply to the availability of these data, which were used
235 under license for the current study, and so are not publicly available. Data are however available
236 from the authors upon reasonable request and with permission of the ZuchtData EDV-
237 Dienstleistungen GmbH.

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245 **Authors' contributions**

246 BKS edited data, performed GWAS and heterogeneity analysis. TS edited data and performed
247 some statistical analyses. CEG contributed to writing of the manuscript. HS edited raw data and
248 contributed to writing of the manuscript. MF calculated pairwise linkage disequilibrium. JS
249 provided the concept of the study, performed some statistical analyses and significantly
250 contributed to writing of the manuscript. All authors read and approved the final manuscript.

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253 **References**

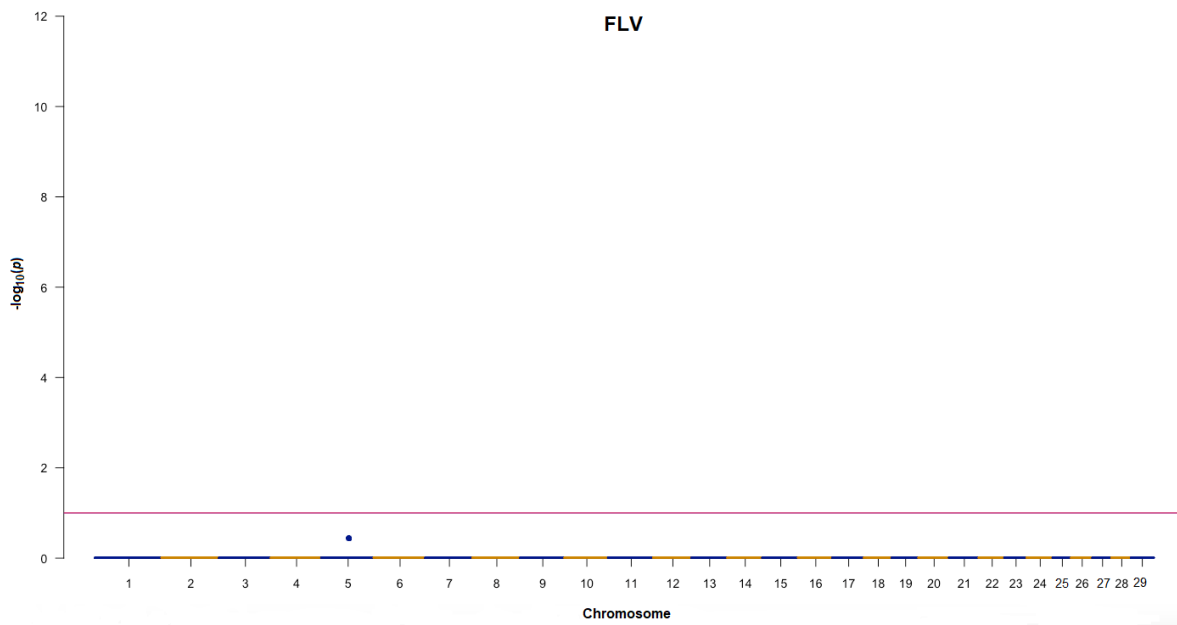
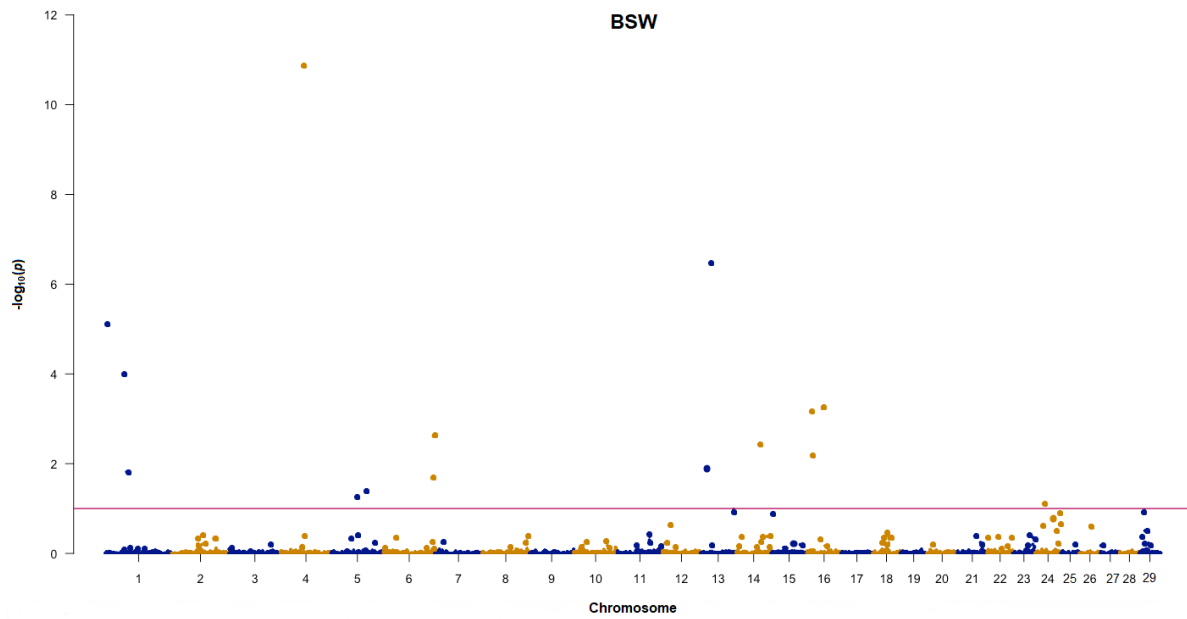
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313 Table 1 Summary of GWAS results.

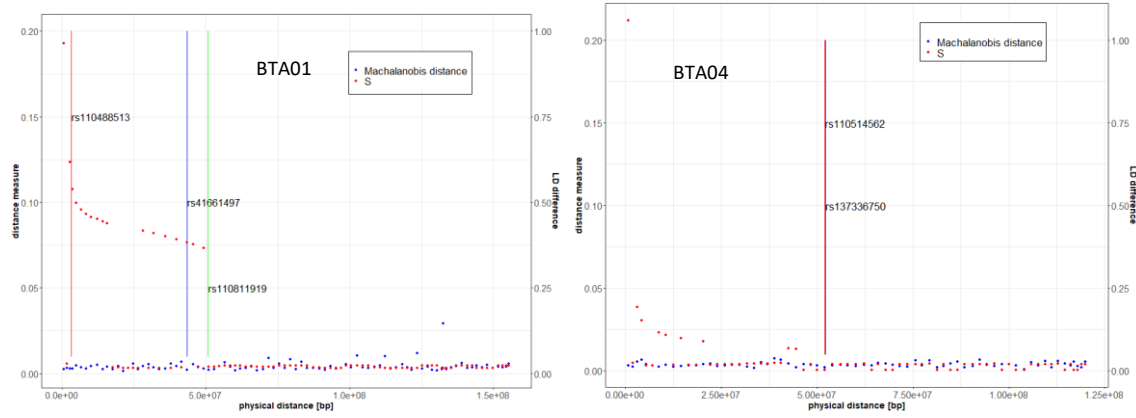
Position name	Closest Gene(s)	QTL	Effect	Increasing allele	FDR – effect	FDR – distance	S	P – frequency ratio
1:3,303,269 rs110488513	Intergenic between MIS18A and HUNK		0.018	A	0.000008		1.0	0.163591 0.391039
1:43,542,488 rs41661497	Intergenic between DCBLD2 and COL8A1		0.012	G	0.000106		1.0	0.152780 0.229225
1:50,767,507 rs110811919	Intergenic upstream of CBLB	Hindquarter proportions (7124)	0.007	G	0.015146		1.0	0.152780 0.007376
4:52,028,036 rs110514562	Intergenic between CAV2 and TES		0.029	A	<10 ⁻⁶		1.0	0.002094 <10 ⁻⁷
4:52,079,221 rs137336750	Intergenic between CAV2 and TES		0.029	G	<10 ⁻⁶		1.0	0.002094 <10 ⁻⁷
5:61,220,624 rs41590733	Intergenic upstream of NEDD1	Rump conformation (3422, 3424, 1563, 20622)	0.010	A	0.052989		1.0	0.005398 0.030671
5:81,769,685 rs109268584	Intergenic between CCDC91 and PTHLH		0.008	A	0.043017		1.0	0.005398 0.008493
6:114,116,280 rs110962969	Intron of SORCS2		0.010	C	0.002456		1.0	0.006265 0.381230
13:13,590,662 rs110792762	Intergenic upstream of CELF2		0.008	G	0.012588		1.0	0.451395 0.003215
13:23,590,146 rs110989397	Intergenic between SPAG6 and PIP4K2A	Rump angle (3429)	0.019	A	<10 ⁻⁶		1.0	0.006076 0.019040
14:55,768,446 rs110534995	Intergenic between TMEM74 and EMC2		0.010	A	0.003962		1.0	0.023641 0.153087
16:12,125,227 rs29024589	Intergenic between CDC73 and GLRX2		0.010	T	0.000676		1.0	0.021476 0.394140
16:12,280,122 rs110843300	Intergenic upstream of UCHL5		0.008	G	0.006398		1.0	0.021475 0.398872
16:36,037,389 rs41579631	Intergenic between RGS7 and ENSBTAG00000009943		0.007	G	0.000592		1.0	0.026922 <10 ⁻¹¹⁵
24:24,273,191 rs136424124	Intergenic between CCDC178 and KLHL14		0.002	A	0.082299		1.0	0.006277 0.111094

315 Figure 1 Manhattan plots for Braunvieh (BSW) and Fleckvieh (FLV). Red line marks down
316 the significant level of 10 %.

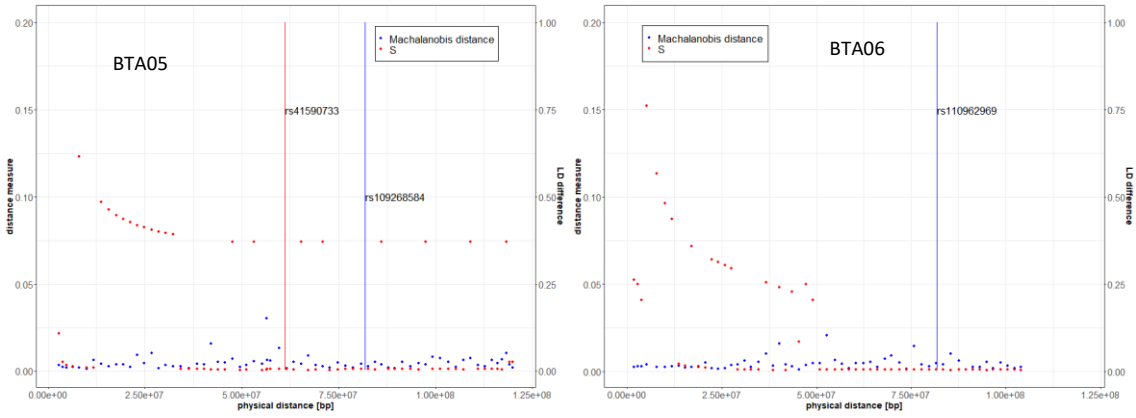


321 Figure 2 LD decay and Machalanobis distance for all chromosomes containing significant
322 SNPs.

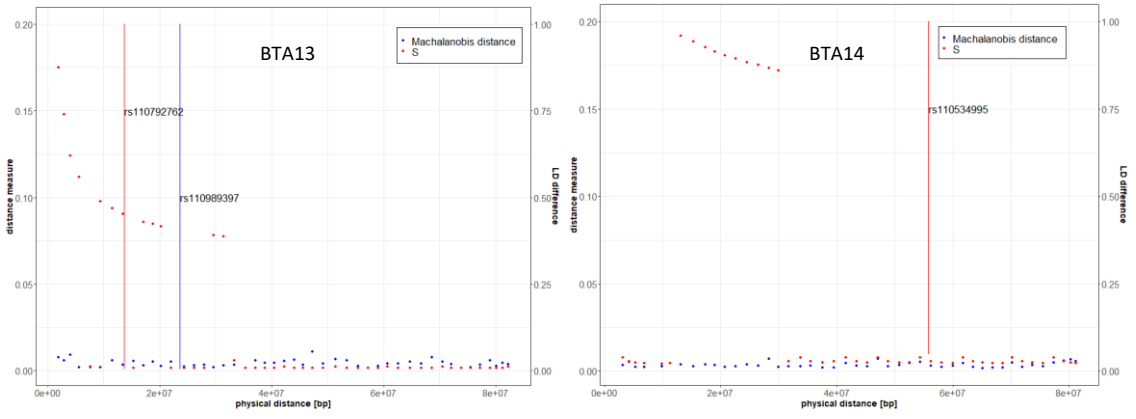
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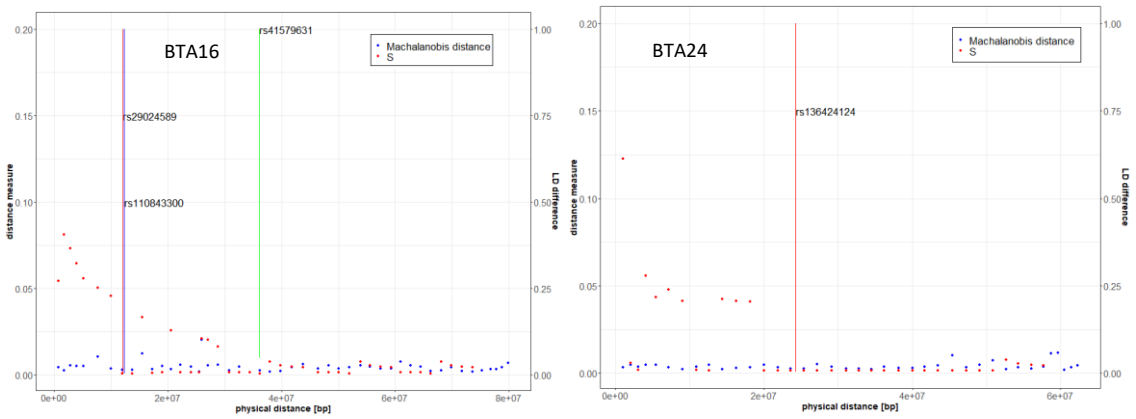
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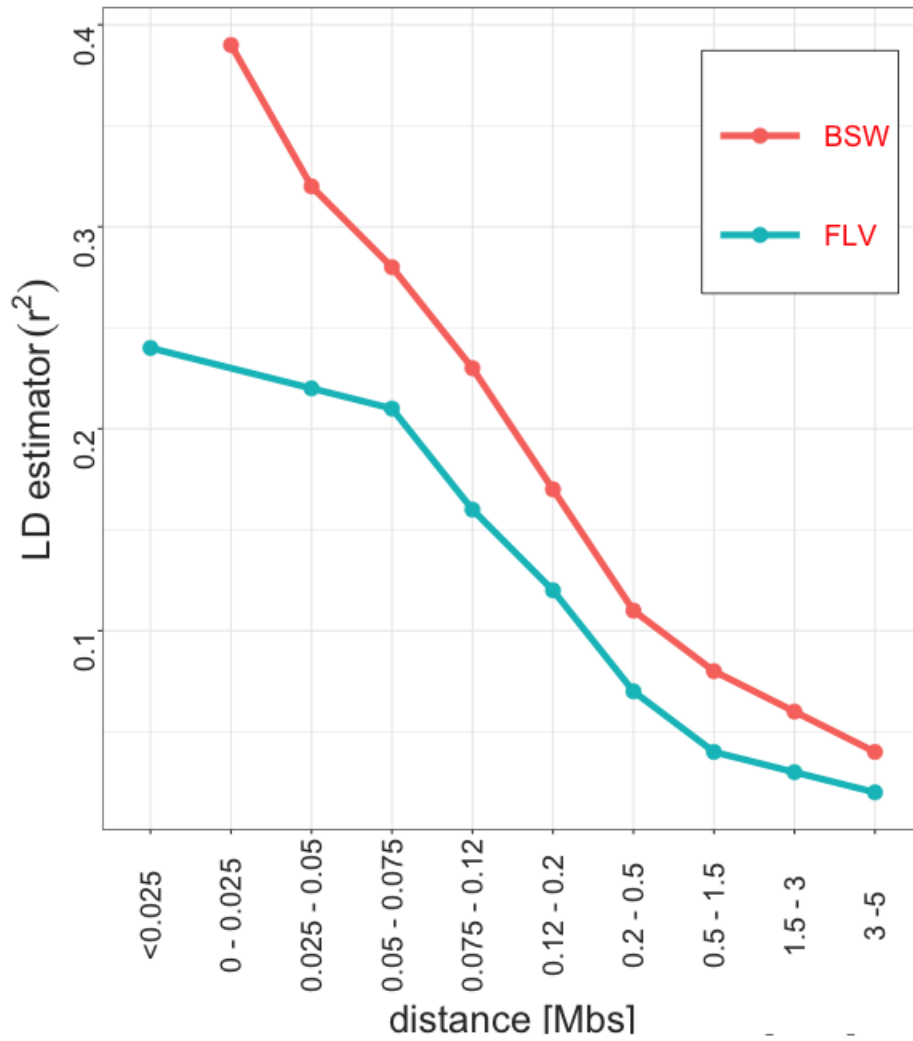
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326



327 Figure 3 Differences in genetic architecture between breeds, expressed by LD patterns across
328 the whole genome.



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