

doi:10.1093/jas/skaa151 Advance Access publication May 4, 2020 Received: 23 January 2020 and Accepted: 25 A

Received: 23 January 2020 and Accepted: 25 April 2020 Animal Genetics and Genomics

ANIMAL GENETICS AND GENOMICS

Breed- and trait-specific associations define the genetic architecture of calving performance traits in cattle

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Abstract

Reducing the incidence of both the degree of assistance required at calving, as well as the extent of perinatal mortality (PM) has both economic and societal benefits. The existence of heritable genetic variability in both traits signifies the presence of underlying genomic variability. The objective of the present study was to locate regions of the genome, and by extension putative genes and mutations, that are likely to be underpinning the genetic variability in direct calving difficulty (DCD), maternal calving difficulty (MCD), and PM. Imputed whole-genome single-nucleotide polymorphism (SNP) data on up to 8,304 Angus (AA), 17,175 Charolais (CH), 16,794 Limousin (LM), and 18,474 Holstein-Friesian (HF) sires representing 5,866,712 calving events from descendants were used. Several putative quantitative trait loci (QTL) regions associated with calving performance both within and across dairy and beef breeds were identified, although the majority were both breed- and trait-specific. QTL surrounding and encompassing the myostatin (MSTN) gene were associated ($P < 5 \times 10^{-8}$) with DCD and PM in both the CH and LM populations. The well-known Q204X mutation was the fifth strongest association with DCD in the CH population and accounted for 5.09% of the genetic variance in DCD. In contrast, none of the 259 segregating variants in MSTN were associated ($P > \times 10^{-6}$) with DCD in the LM population but a genomic region 617 kb downstream of MSTN was associated ($P < 5 \times 10^{-8}$). The genetic architecture for DCD differed in the HF population relative to the CH and LM, where two QTL encompassing ZNF613 on Bos taurus autosome (BTA)18 and PLAG1 on BTA14 were identified in the former. Pleiotropic SNP associated with all three calving performance traits were also identified in the three beef breeds; 5 SNP were pleiotropic in AA, 116 in LM, and 882 in CH but no SNP was associated with more than one trait within the HF population. The majority of these pleiotropic SNP were on BTA2 surrounding MSTN and were associated with both DCD and PM. Multiple previously reported, but also novel QTL, associated with calving performance were detected in this large study. These also included QTL regions harboring SNP with the same direction of allele substitution effect for both DCD and MCD thus contributing to a more effective simultaneous selection for both traits.

Key words: calving difficulty, genome-wide association, mortality, pleiotropy, sequence, single-nucleotide polymorphism

Introduction

The association between the extent of assistance required at calving and subsequent performance has been extensively documented in both dairy (Dematawewa and Berger, 1997; Berry et al., 2007; Eaglen et al., 2011) and beef (Drennan and Berry, 2006)

cattle; similarly, the association between perinatal mortality (PM) and subsequent performance has been reported in several cattle populations (Berry et al., 2007). Calving dystocia has been associated with reduced subsequent reproductive performance (Dematawewa and Berger, 1997; Berry et al., 2007; Eaglen et al., 2011),

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Abbreviations

AA	Angus
BTA	Bos taurus autosome
CH	Charolais
DCD	direct calving difficulty
EBV	estimated breeding value
ERC	effective record contribution
HD	high density
HF	Holstein-Friesian
ICBF	Irish Cattle Breeding Federation
IDB	Irish Dairy and Beef
LCORL	ligand-dependent nuclear receptor
	corepressor-like
LD	linkage disequilibrium
LM	Limousin
MAF	minor allele frequency
MCD	maternal calving difficulty
MSTN	myostatin
NCAPG	non-SMC condensin I complex
	subunit G
PARM1	prostate androgen-regulated
	mucin-like protein 1
PDK4	pyruvate dehydrogenase kinase 4
PLAG1	pleomorphic adenoma gene 1
PM	perinatal mortality
PRR16	proline rich 16
QTL	quantitative trait loci
SCN4B	sodium voltage-gated channel beta
	subunit 4
SLC9A2	solute carrier family 9 member A2
SNP	single-nucleotide polymorphism
WGS	whole-genome sequence
ZEB2	zinc finger E-box-binding homeobox 2
ZNF613	zinc finger protein 613.

reduced milk production (Dematawewa and Berger, 1997; Berry et al., 2007; Eaglen et al., 2011), greater body condition score loss postcalving (Berry et al., 2007), as well as compromised health (Berry et al., 2007) in dairy cattle; the association in dairy cattle between dystocia and both calf performance and survival has also been documented (Peeler et al., 1994; Meyer et al., 2001; Berry et al., 2007). While a reduction in performance obviously affects farm profit, the periparturient period is also known to be a very stressful period in the lifetime of, not only the mature female, but also the progeny (Eaglen et al., 2011). Hence, reducing the extent of assistance required during the birthing process, as well as reducing PM rates, should not only improve herd profit but also the well-being status of the herd. Therefore, all potentially useful strategies to ameliorate these repercussions should be thoroughly investigated. Breeding program, with or without the exploitation of genomic information, as an approach to reduce the reliance on calving assistance or reducing the risk of PM is one such strategy.

The extent of genetic variability underlying phenotypic differences in calving dystocia has been well-documented in both dairy (Eaglen and Bijma, 2009) and beef cattle (Mujibi and Crews, 2009; Crowley et al., 2011); similarly, genetic variation in PM in cattle is also known to exist (Crowley et al., 2011; Ring et al., 2018). While periparturient management strategies are known to influence the risk of calving difficulty, direct heritability estimates for calving dystocia in cattle have been documented to range from 0.08 to 0.24 (Eaglen and Bijma, 2009; Mujibi and Crews, 2009; Crowley et al., 2011) with reported maternal heritability estimates ranging from 0.02 to 0.10 (Eaglen and

Bijma, 2009; Mujibi and Crews, 2009; Crowley et al., 2011). Direct and maternal heritability estimates for PM in cattle have been reported to range from 0.01 to 0.03 (Crowley et al., 2011; Ring et al., 2018). Hence, being able to explain the additive genetic variation in both calving difficulty (i.e., direct and maternal) and PM through genomic markers, or the causal mutations themselves, could not only aid breeding schemes attempting to reduce the incidence of both, but such information could also facilitate optimized mating programs as a tool to minimize the likelihood of requiring assistance at calving and/or of PM.

The objective of the present study was to use imputed wholegenome sequence (WGS) data from a large population of cattle in an attempt to locate regions of the bovine genome thought to harbor genes and polymorphisms contributing to the genetic variability in both direct (DCD) and maternal calving difficulty (MCD) as well as PM. Of particular interest in the present study was the exploitation of data from multiple breeds in an attempt to quantify if the associated regions persisted across breeds, but also if commonalities existed between the detected genomic regions for the three traits investigated thus aiding in the resolving of known genetic antagonisms (Crowley et al., 2011).

Materials and Methods

The data used in the present study originated from a preexisting database managed by the Irish Cattle Breeding Federation (**ICBF**). Therefore, it was not necessary to obtain animal care and use committee approval in advance of conducting this study.

Phenotypes

In Ireland, calving difficulty is subjectively scored by producers on a linear scale of 1 to 4 at the time of birth, where 1=no calving assistance; 2=slight assistance (assistance by one person, without needing to use a calf puller); 3=considerable assistance (assistance by one person using a calf puller or more than one person); 4=veterinary assistance (including caesarean). Perinatal mortality is also recorded as a binary variable by producers, and indicates whether the calf died within a 24-h period after birth.

Estimated breeding values (EBVs) for DCD and MCD, as well as direct PM with their respective reliability estimates were obtained from the ICBF database from the December 2018 national genetic evaluation. The DCD and MCD breeding value estimates were generated in a direct-maternal multibreed genetic evaluation, whereas PM estimates were generated in a univariate multibreed model (i.e., no maternal genetic component); the heritability of each trait in the respective models was 9% for DCD, 2% for MCD, and 2% for PM. Of the animals with generated EBVs, only purebred (i.e., ≥87.5% of a single breed) genotyped sires of any of the four numerically largest breeds were retained for analysis; Angus (AA), Charolais (CH), Holstein-Friesian (HF), and Limousin (LM). The effective record contribution (ERC) of each sire, taking into consideration the genotyped animals, was estimated using the Harris and Johnston (1998) method; only animals with an ERC \geq 1 were retained for use in the present study. Deregression of the EBVs was completed using the secant method with a full animal model pedigree file. The number of animals per trait after edits is detailed in Table 1, as well as the median ERC.

Genotype data

Genotypes of 60,464 sires with calving EBVs were imputed to WGS as part of a larger data set of 638,662 genotyped animals from multiple breeds as detailed previously by Purfield et al. (2019a). All animals

Table 1. The number of animals included	per trait and breed
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Trait	A A	СН	HF	LM	All
Ifait	AA	CH	HF	LIVI	All
Direct calving difficulty					
Number of animals	8,304	17,175	18,191	16,794	60,464
Median ERC	7.31	13.38	2.23	13.57	8.40
Number of SNP after edits	14,778,952	16,091,418	14,331.494	16,144,687	16,713,783
Maternal calving difficulty					
Number of animals	7,013	11,792	16,265	12,683	47,753
Median ERC	2.79	3.49	3.79	4.12	3.56
Number of SNP after edits	14,799,631	16,131,655	14,380,942	16,1814,561	16,498,270
Direct mortality					
Number of animals	8,104	16,446	18,474	15,873	58,897
Median ERC	6.76	10.87	3.35	11.32	7.48
Number of SNP after edits	14,791,231	16,088,788	14,348,548	16,148,558	16,487,386.

included in the present study were genotyped on a variety of genotype panels including the Illumina High Density (HD; n = 3,042; 777,962 SNP), Illumina Bovine SNP50 (n = 1,094; 54,001 SNP), the custom Irish Dairy and Beef (IDB) V1 (n = 14,001; 16,622 SNP), the IDBV2 (n = 26,284; 16,223 SNP), or the IDBV3 (n = 16,043; 52,445 SNP) genotype panels. All 638,662 genotyped animals had a call rate $\ge 90\%$ and only autosomal SNP, SNP with a known chromosome and position on UMD 3.1, and only SNP with a call rate $\ge 90\%$ were retained within each panel.

Prior to imputation to WGS, all genotyped animals of the larger 638,662 genotyped data set were first imputed to HD using a two-step approach in FImpute (Sargolzaei et al., 2014); this involved imputing all IDB-genotyped animals to the Bovine SNP50 density and subsequently imputing all resulting genotypes, including the Bovine SNP50 genotypes, to HD using a multibreed reference population of 5,504 HD-genotyped animals. Imputation of all 638,662 HD-imputed animals to WGS was then undertaken using a reference population of 2,333 Bos *taurus* animals of multiple breeds from Run6.0 of the 1000 Bull Genomes Project. Imputation of the HD genotypes to WGS was achieved by firstly phasing all 638,662 HD-imputed animals using Eagle (Loh et al., 2016; version 2.3.2) and subsequently imputing all animals to WGS using minimac3 (Das et al., 2016).

Regions with possible poor WGS imputation accuracy (n = 687,137 SNP) were identified using a data set of 147,309 verified parent-progeny relationships from the 638,662 genotyped data set as previously described by Purfield et al. (2019a) and subsequently removed from the data set. In addition, all SNP with a minor allele frequency (MAF) < 0.005 across all animals were removed for the multibreed analysis and all SNP with a MAF < 0.005 within each breed were removed for the respective within-breed analysis. The number of SNP that remained for analysis within each breed is detailed in Table 1.

Genome-wide association analyses

Whole-genome association analyses were performed within each breed separately, as well as in a data set of all breeds combined, using an animal linear mixed model in Wombat (Meyer and Tier, 2012). The VanRaden Method I (2008) was used to create a genomic relationship matrix based on just the imputed autosomal SNP from the edited HD panel (n = 642,153 SNP). All imputed sequence SNP, scored as 0, 1, or 2, were included individually as a fixed effect covariate in the model 1 at a time. Breed was also included as a fixed effect for the multibreed analyses. Each dependent variable was also weighted using the approach outlined by Garrick et al. (2009);

$$w_i = \ rac{1-h^2}{\left[c+rac{1-r_i^2}{r_i^2}
ight]h^2}$$

where w_i is the weighting factor of the ith deregressed EBV, h^2 is the heritability estimate for the trait in question, r_i^2 is the reliability of the ith deregressed EBV, and c is the proportion of genetic variance not accounted by the SNP and set at 0.9 for analyses thus allowing each SNP to attribute up to 10% of the genetic variance. Test statistics for all SNP were obtained and converted into their corresponding P-values. The genomic inflation factor was estimated within each breed and ranged from 0.99 in the AA population to 1.08 in the LM population, whereas the across-breed inflation factor ranged from 0.98 for MCD to 1.07 for DCD. A genome-wide SNP significance threshold of P \leq 5 \times 10⁻⁸ and a suggestive threshold of P \leq 1 \times 10⁻⁶ were applied to each analysis (Pe'er et al., 2008). The proportion of the genetic variance attributable to individual SNP was calculated as $2pqa^2/\sigma^2$, where p was the major allele frequency, q was the MAF, a was the estimated allele substitution effect, and σ^2 was the genetic variance for the respective trait.

Defining quantitative trait loci

The definition of a quantitative trait loci (QTL) was undertaken as outlined in detail by Purfield et al. (2019a) and McGovern et al. (2019). All SNP surpassing the suggestive significance threshold $(P \le 1 \times 10^{-6})$ were used for defining QTL regions associated with each trait separately. To estimate the QTL start and end positions, all SNP within a 5-Mb window and in strong linkage disequilibrium (LD) (r^2 of ≥ 0.7) with each suggestively associated SNP were considered to be part of the QTL. Overlapping QTL were merged together and considered as the same QTL. To limit the number of false-positive QTL, a minimum of two suggestively associated SNP had to be present in the QTL region. Genes within and overlapping each QTL were identified using NCBI map viewer (http://www.ncbi.nlm.nih.gov/mapview) and Ensembl (http://ensemble.org) based on the bovine UMD 3.1 build. Candidate genes were chosen from the QTL region based on previous literature and their biological function. If no gene resided in the QTL region, genes within 250 kb of the start and end position of the QTL were considered as putative candidate genes. Previously reported cattle QTL were obtained from the animal (http://www.animalgenome.org/cgi-bin/QTLdb/index). OTLdb In addition, to identify QTL present in more than one breed, each chromosome was split into 500-kb windows and each window that contained a SNP with a $P \le 1 \times 10^{-4}$ present in two or more breeds was considered a putative across-breed QTL. A similar approach was used to detect QTL common to all three calving traits. This threshold was previously applied by Tenghe et al. (2016) when detecting across-trait QTL thus enabling the identification of putative across-breed genomic regions with less stringency.

Results

Genotypes from more than 40 million sequence variants were imputed for the 60,464 sires from the four cattle breeds which represented a total of 5,866,712 calving events from their descendants used in the estimation of their EBVs.

Direct calving difficulty

As more genome-wide significant SNP were associated with DCD than both MCD and PM, only QTL regions containing genome-wide significant ($P < 5 \times 10^{-8}$) SNP are listed in Table 2 for DCD. The number of significant QTL regions associated with DCD differed per breed ranging from 2 QTL regions in the HF and LM populations to 17 in the CH population (Table 2). The strongest association with DCD resided on Bos taurus autosome (BTA)2 in the CH population where a QTL region encompassing myostatin (MSTN) contained 146 significantly associated (P < 5 × 10-8) SNP (Figure 1B). The fifth strongest association within this QTL on BTA2 in the CH population was the well-known Q204X stop-gain mutation within MSTN which had a P-value of 1.42 × 10-14 and explained 5.09% of the genetic variance in DCD. In total, 44 MSTN-annotated SNP within this QTL spanning from 58.72 to 67.21 Mb were significantly associated ($P < 5 \times 10^{-8}$) with DCD in the CH population. Overall, there were 26 CH animals within the edited data set that were homozygous for the T allele at the Q204X mutation and these had greater (P < 0.001) calving difficulty (mean DCD EBV 0.482, SD 0.461) than the 15,452 noncarriers of the Q204X mutation (mean DCD EBV of 0.182, SD 0.214) (Supplementary Figure 1). Similar to CH, the strongest association within the LM population was also on BTA2, but it was 617 kb downstream of the MSTN. Indeed none of the 259 SNP within MSTN segregating in the LM were associated (P > 1 × $10^{\text{-6}}$) with DCD in the LM population; the A allele of the F94L (rs11065568) mutation in MSTN (P = 1.54×10^{-4}) had a positive allele substitution effect on DCD.

The strongest association in the AA population for DCD was on BTA6 in the QTL spanning from 88.67 to 88.94 Mb which contained the GC gene which encodes for the vitamin D-binding protein; no SNP within GC was significantly or suggestively associated with DCD with the strongest association (rs211646036) having a P-value of 5.73 × 10⁻⁴. The strongest association within the HF population was rs381577268 (P-value = 1.63×10^{-9}), a downstream variant of zinc finger protein 613 (ZNF613) located in a QTL positioned between 58.14 and 58.41 Mb on BTA18; the SNP accounted for 1.32% of the genetic variance in DCD (Table 2). Within the edited data set, 19 animals were homozygous for the T allele of the rs381577268 SNP and these animals had greater (P < 0.001) EBVs (i.e., worse) for DCD (mean EBV 0.118, SD 0.148) than both heterozygous (n = 865; mean EBV 0.002, SD 0.102) and homozygous CC sires (n = 17,307; mean EBV -0.056, SD 0.069) (Supplementary Figure 2). The allele frequency of the T allele was, nonetheless, low within the HF population with a frequency of 0.025. Similar low frequencies were also observed in the beef populations where the T allele had a frequency of 0.007 and 0.005 in the AA and CH populations, respectively, and below the MAF threshold for inclusion in the LM population (MAF = 0.002). Despite rs381577268 segregating within each of the beef populations investigated, neither it, nor any other variant within ZNF613, was associated with DCD in these breeds. However, several SNP from 57.88 to 57.98 Mb in the CH population and from 58.01 to 58.23 Mb in the LM population did have P-values ranging from 5.64×10^{-5} to 4.81×10^{-3} for DCD.

Another plausible QTL was also identified in the HF population on BTA14 in the region of 24.48 to 28.12 Mb encompassing the pleomorphic adenoma gene 1 (PLAG1) gene. In total, 58 SNP within this QTL were significantly associated with DCD and the most significant SNP, rs134489103, was only 53 kb upstream of PLAG1; 21 PLAG1-annotated SNP were segregating in the HF population and their P-values ranged from 8.81×10^{-7} to 9.01×10^{-6} . This QTL on BTA14 is most likely a dairy-specific QTL as none of the 835 (CH) to 1219 (AA) segregating variants within or 250 kb up/downstream of PLAG1 were associated with DCD in the beef populations analyzed (P > 0.023).

The across-breed analyses involving all 60,464 sires with breed fitted as a fixed effect for DCD identified 115 significant and 289 suggestively associated SNP. Twenty-seven QTL across 18 BTAs where identified, of which the strongest QTL association was located on BTA2 encompassing MSTN (Supplementary Table 1). Multiple novel QTL associations were identified, but the allele frequency of the lead variant within these QTL tended to be quite low; of the 27 QTL identified, the lead variant within 14 of the QTL had a MAF < 0.01. As genomic regions rather than individual SNP may be influencing DCD across all breeds, overlapping 500kb windows that contained at least one SNP with a P-value $< 1 \times 10^{-4}$ within each breed were identified (Figure 2). The majority of the 500-kb windows harboring a SNP with a P-value $< 1 \times 10^{-4}$ were unique to a single breed and no window was significant ($P < 1 \times 10^{-4}$) in all four breeds for DCD. Despite this, 31 windows (out of a total of 2,150 unique windows across all breeds) were shared between the CH, LM, and HF populations; these were spread across multiple autosomes (BTA1, 3, 4, 5, 7, 9, 10, 15, 18, 23, 24, and 28) and include plausible candidate genes such as INSIG1 on BTA4, NUDT12 on BTA7, GMDS on BTA23, and GRID which encodes GluD1 on BTA28. In addition, a further 12 genomic windows across 11 autosomes were shared between the AA, LM, and HF populations and included the candidate genes zinc finger E-box-binding homeobox 2 (ZEB2) on BTA2, PDCD5 on BTA18, and ARHAP5 and AKAP6 on BTA21. Although a large proportion of windows were shared between the AA and CH populations, and between the CH and LM populations, no windows were shared among all three beef breeds.

Maternal calving difficulty

Genome-wide significant associations with MCD were detected in all three beef breeds but not in the HF population. The AA population had the greatest number of QTL associated with MCD; in total, 12 QTL regions contained SNP that were suggestively $(P < 1 \times 10^{-6})$ associated with MCD (Table 3). The strongest association within the AA population was an intergenic variant, rs798343691, on BTA5 (P = 1.66×10^{-9}) (Figure 3), although no other SNP in LD ($r^2 \ge 0.7$) with this SNP in the AA was suggestively associated with MCD and, as such, it was not considered a putative QTL region. The QTL region containing the strongest association within the AA population was on BTA12 spanning from 56.67 to 56.68 Mb where one SNP was significantly associated with MCD and four others were suggestively associated with MCD; no putative candidate gene within 250 kb upstream or downstream was identified. Several intronic variants within the candidate gene prostate androgen-regulated mucin-like protein 1 (PARM1) were suggestively associated with MCD in the QTL from 91.60 to 91.63 Mb on BTA6 but, of the two segregating missense variants within PARM1 (rs111027720 and rs207780467) captured within the data set, neither were significant (P-value \geq 0.1). In addition, a QTL region surrounding the well-known statureassociated genes, non-SMC condensin I complex subunit G (NCAPG) and ligand-dependent nuclear receptor corepressorlike (LCORL), was also associated with MCD in AA, although none of the 241 and 553 segregating SNP actually within each

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$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	$rs384824032^{a}$	3241507	Т	0.240	6.09×10^{-9}	I	‡	0	N/A
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	$rs210421829^{a}$	3899839	A	0.457	5.53×10^{-9}	I	‡	0	N/A
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	$rs482419628^{a}$	6630781	ტ	0.054	1.09×10^{-22}	ł	‡	6	MSTN
$ \begin{array}{llllllllllllllllllllllllllllllllllll$	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	$rs799943285^{b}$	6808074	ტ	0.058	4.95×10^{-21}		+	1	SLC40A1 [†]
4 118379848 118950619 12 rs377892980 ^d 11837395 T 0.008 2.30×10^4 - + 4 7 111633395 111633395 111633395 111633395 111633395 - ++ 0 8 99124662 99444787 4 9124662 0 - ++ 0 115 80425653 80425650 ² 28425672 C 0.008 5.81 \times 10^{*11} -+ + 1 20 2763982 28292172 14 27888972 C 0.008 5.81 \times 10^{*11} -+ + 1 20 4075568 41092503 7 11639445 A 0.007 5.81 \times 10^{*1} -+ + 1 1 21 249<708	$ \begin{array}{llllllllllllllllllllllllllllllllllll$	$rs798702162^{a}$	7609482	ტ	0.051	8.61×10^{-13}		‡	7	COL3A1
7 11163335 111633100 2 1533385 111633100 2 1533385 11163310 2 1533385 111633355 111633355 111633355 111633355 111633355 111633355 111633355 111633355 111633355 111633355 111633355 111633355 111633355 111633355 111633355 111633355 111633355 111633355 111633355 111633355 11163355 11163355 11163355 11163355 11163355 11163355 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 111174 1111174 1111174	71116333951116391002rs383889912466299444787415 80425653 80425659 2rs383820 27639892 28292172 1420 27639892 28292172 1420 40735668 41092503 7rs717221 988356 1889045 1122 988356 1889045 1123 21839465 21839742 4rs460523 20541765 14 rs2085328 20486363 22497098 2rs208528 45199788 249761 rs2085rs314614 24847507 27124281 262 rs314518 58141989 58407174 8rs381519 14707689 5 rs321352 55769372 552 rs211372 6692259 6837692 94 rs13775 55069259 68376925 3 rs11157 75765756 552 rs211367 75765756 552 rs211377 7576	$rs377892980^{d}$	118932719	υ	0.990	2.80×10^{-8}	I	‡	4	RNF32 ⁺ , LMBR1, NOM1
8 9912462 9944787 4 \cdot 9912462 C 0.988 3.04 × 10 ⁴ ++ 2 15 80425653 80425653 2 1542805962 ^d 80425653 C 0.008 5.81 × 10 ⁻¹¹ 1 - 1 - 1 - - 1 1 1 1 1 1 1 1 1 1 1 1 1 1	8 99124662 99444787 4 15 80425653 80425659 2 rs4280 20 27639892 28292172 14 rs7172 20 27639892 28292172 14 rs7172 20 207639892 28292172 14 rs7172 21 388356 41092503 7 rs7172 22 988356 1889045 11 rs7172 23 21839465 21839742 4 rs4605 23 21839465 21839742 4 rs4605 23 21839465 21839742 4 rs4605 25 22436613 20541765 14 rs2106 28 20486564 24389785 3 rs2106 28 24369748 2447507 27124281 262 rs31461 14 24847507 27124281 262 rs3146 rs32106 28 581401999 584407174 8 rs32104	rs383826293ª	111633395	T	0.008	2.49×10^{-8}	I	‡	0	N/A
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23 21839465 21839742 4 1346657227^{16} 21839717 T 0.007 4.87×10^{-8} - ++ 1 25 22436613 22497098 2 13385805439° 22458425 T 0.005 6.56×10^{-9} NA + 1 28 20486363 20541765 14 1220859752° 22056931 C 0.012 3.23×10^{-9} - - + 1 28 24368554 24389785 3 128710° 4535972° 20566931 C 0.012 3.23×10^{-9} - - + 1 28 24368554 24388748 24 14774662° 2435371° 45353728° 22453552° 22456874 A 0.986 8.80×10^{-11} + - 17 14 24847507 27124281 262 18381577268° 58141989 T 0.025 1.51×10^{-9} + - 17 1	23 21839465 21839742 4 25 22436613 22497098 2 28 20486363 20541765 14 28 20486363 20541765 14 28 20486363 20541765 14 28 20486363 20541765 14 28 20486363 20541765 14 28 24389785 3 24 28 45109768 45388748 24 14 24847507 27124281 262 18 58141989 58407174 8 1 147046621 147071689 5 2 5751482 5769372 52 2 669259 68337692 94 5 55007482 55162856 3 7 57562856 551662856 3	а.	1889045	A	0.007	3.55×10^{-9}	NA	I	∞	SLC4A7
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28 20486363 20541765 14 $rs20859725^a$ 20506931 C 0.012 3.23×10^9 $$ $$ 0 28 24368554 2438785 3 $rs210629062^a$ 24368554 C 0.035 3.61×10^8 $++$ $+$ $ 0$ 28 45109768 45338748 24 $rrs476163771^a$ 45359784 A 0.986 8.80×10^{-11} $++$ $ 0$ 14 24847507 27124281 262 $rrs134489103^d$ 24953585 T 0.968 8.80×10^{-11} $++$ $ 0$ 18 58141989 58407174 8 $rrs381577268^d$ 58141989 T 0.096 8.80×10^{-11} $++$ $ 0$ 1 147046621 147071689 5 $rrs1381577268^d$ 58141989 T 0.002 3.21×10^{-9} $+$ $+$ 1 2 5751482 5769372 52 $rrs211366913^d$ 5751482 576976^{-10} $ +$ $+$ 1 2 5751482 5769372 52 $rrs211366913^d$ 5751482 576976^{-10} $ +$ $+$ 1 2 6692259 6837692 94 $rrs13771945^a$ 5751482 5769×10^{-9} $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ $+$ <	28 20486363 20541765 14 28 24368554 24389785 3 28 24368554 24389785 3 28 45109768 45388748 24 14 24847507 27124281 262 18 58141989 58407174 8 1 147046621 147071639 5 2 5751482 5769372 52 2 6692259 6837692 94 5 851007482 85162856 3 7 5756555 55009456 3	$rs385805439^{\circ}$	22458425	F	0.005	6.56×10^{-9}	NA	+	1	$RBBP6^{\dagger}$
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5751482576937252rs211366913d5751482G0.048 6.19×10^{-9} -++16692259683769294rs137213740a6696029T0.9097.69 \times 10^{-11}++185007482851628563rs111563671c85162856T0.9722.29 \times 10^{-8}++157666736582981624rs133717945a57666736A0.0675.43 \times 10^{-9}++1239286324613845rs378942567a42372750G0.0033.57 \times 10^{-8}++1640647337423727502rs378942567a42372750G0.0033.57 \times 10^{-8}++16	5751482 5769372 52 6692259 6837692 94 85007482 85162856 3 57662736 50000450 3		147046683	Ċ	0.00	3.21×10^{-9}	I	+	1	$SLC19A1^{\dagger}$
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$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$		$rs111563671^{\circ}$	85162856	Т	0.972	2.29×10^{-8}	+	+	1	KRAS [†]
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40647337 42372750 2 $rs378942567^a$ 42372750 G 0.023 3.57×10^8 – ++ 16	2392863 2461384 5	$rs111672216^{a}$	2461384	Ŧ	0.007	6.67×10^{-10}	+++++	I	0	N/A
		rs378942567ª	42372750	ტ	0.023	3.57×10^{-8}	I	‡	16	PPYR1

effect size for MCD, i.e., a SNP effect size in the top 20%, - represents opposing SNP effect and -- represents opposing SNP effect directions that have a large SNP effect size. N/A (Non-applicable) was if the SNP wasn't present in that dataset due to allele frequency. SNP effect PM similar to MCD but represents the SNP effects for PM. No. of genes is the number of genes in the QTL. If no gene was present in the QTL the nearest functional candidate gene within 250 kb was chosen. ^aintergenic, ^bintronic, ^cupstream gene variant, ^ddownstream gene variant, ¹gene in which the most significant SNP within the QTL was identified in.

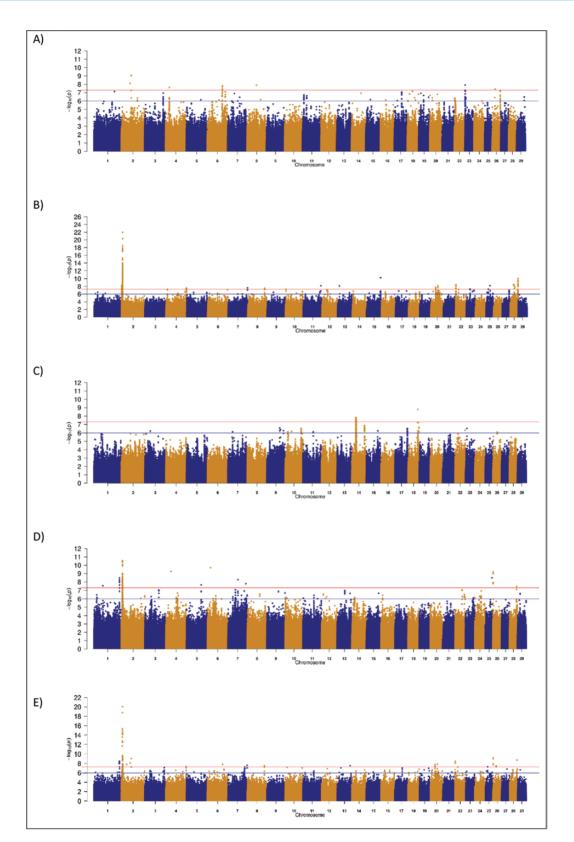


Figure 1. Manhattan plots for direct calving difficulty in (A) Angus, (B) Charolais, (C) Holstein Friesian, (D) Limousin, and (E) across all breeds. The red line indicates the genome-wide significance threshold of $P \le 5 \times 10^{-8}$ and the blue is the suggestive significance threshold of $P \le 1 \times 10^{-6}$.

of the genes, respectively, were suggestively associated with MCD. The strongest association with MCD in the CH population was on BTA11, where 11 variants within a QTL from 92.40 to

92.47 Mb were significantly associated with MCD. No candidate genes were identified within the QTL boundaries although both GGTA1 and DAB2IP were located within 250 kb of the QTL. The

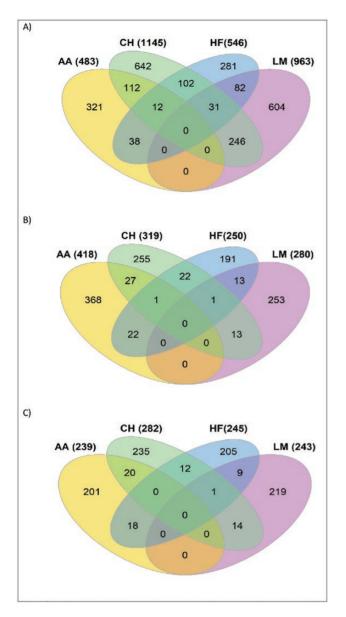


Figure 2. The number of overlapping 500-kb windows that contained a SNP with a P-value < 10^{-4} for (A) direct calving difficulty, (B) maternal calving difficulty, and (C) PM. Yellow represents Angus (AA), green is Charolais (CH), blue is Holstein-Friesian (HF), and pink is Limousin (LM).

most significant variant within this QTL (rs380146645) in the CH population explained 3.76% of the genetic variance in MCD but the favorable C allele associated with an improvement in MCD was near-fixation (allele frequency = 0.994). In total, three QTL were identified to be suggestively associated with MCD in the HF population, all of which were located on BTA26; only one QTL on BTA1 was identified in the LM population (Table 3). Plausible candidate genes identified in these QTL include *GLRX3* and DPYSL4 in the HF population and EPHA6 in the LM population.

Only one QTL region associated with MCD was identified in the across-breed analysis of all 47,753 sires; this QTL which was not previously identified in any of the within-breed analyses contained two suggestively associated intronic SNP which were located in solute carrier family 9 member A2 (SLC9A2) on BTA11 (Supplementary Table 1). No genomic window was associated with MCD across all breeds in the window-based analyses. Two separate genomic windows on BTA7 were, however, associated with MCD in the AA, CH, and HF populations and in the CH, HF, and LM populations, respectively; candidate genes within these windows include ENSBTAG0000003086 in the AA, CH, and HF populations and CANX, CBY3, HNRNPH1, and ADAMTS2 in the CH, HF, and LM populations.

Direct PM

Several genomic regions were significantly associated with PM in the CH, HF, and LM populations; no region was significant in the AA (Figure 4). The strongest association for PM was on BTA2 in the CH population, where a QTL spanning from 5.38 to 6.94 Mb encompassing MSTN contained 335 significantly associated SNP; the Q204X stop-gain mutation was the 26th strongest association with a P-value of 3.54×10^{-12} . The frequency of the Q204X T allele was 0.051. Homozygous carriers of the Q204X T allele (n = 22) had a worse (P < 0.0001) EBV for mortality (mean EBV 0.006, SD 0.015) than homozygous (n = 14,844) noncarriers (mean EBV -0.0002, SD 0.013) (Supplementary Figure 3). Aside from MSTN, a putative QTL was also identified on BTA4 in the CH population, where 60 SNP within the plausible candidate gene pyruvate dehydrogenase kinase 4 (PDK4) were suggestively associated with PM (Table 4). In total, three missense variants within PDK4 were captured within the 1000 Bull data set, but only one (rs109626492) had a MAF > 0.005 in the CH population and this SNP was not associated with PM (P = 0.46); of the remaining two missense variants, rs515875763 was fixed for the nondeleterious allele and rs522912661, which had a MAF of 3.04×10^{-5} , was included in a post hoc analysis but was not associated with PM (P = 0.014). The strongest QTL association with PM in the HF population was also on BTA4, albeit 4 Mb further downstream from the end of the identified CH QTL; the nearest gene to this HF QTL was RPA3 located 398 kb upstream and closer to the CH QTL.

In total, four QTL regions were associated with PM across all breeds (Supplementary Table 1); three were located on BTA2 from 5.97 to 7.61 Mb, of which the QTL from 5.97 to 6.63 Mb encompassed MSTN was the strongest association with PM. The across-breed analyses identified a novel QTL on BTA4, downstream of the gene ZNF800 which had not been previously identified in any of the within-breed analyses. Similar to DCD and MCD, no genomic region associated with PM was common to all four breeds and only one 500-kb window was associated with PM in three breeds; this region was located on BTA4 within the CH, HF, and LM populations with PDK4 being the plausible candidate gene. The minimal overlap identified across breeds suggests that a greater proportion of SNP with a P-value < 10^{-4} are breed-specific associations in comparison to those associated with either DCD or MCD.

Direct PM adjusted for calving difficulty

Upon inclusion of the EBV for DCD as a covariate in the statistical model for the association analysis relating PM to each SNP in the CH population, the QTL on BTA2 encompassing MSTN remained associated with PM although it reduced in significance (Supplementary Table 2). In total, five QTL were identified in the CH population using this approach, of which three were previously identified in the PM analysis suggesting that these QTL contribute still to the incidence of PM, irrespective of a difficult calving. Of most interest, however, was the novel QTL on BTA15 from 29.26 to 29.88 Mb which contained four suggestively associated

Breed	BTA	Start	End	No. SNP	Strongest SNP	Strongest SNP position	+ Allele	+ Allele freq	P-value	SNP effect SNP effect DCD PM	SNP effect PM	No. genes	Candidate gene(s)
AA	1	36828793	36860576	2	æ*	36828793	υ	0.008	2.89×10^{-7}	l		1	EPHA3
	1	144496949	144546610	2	$rs479310525^{b}$	144545050	Ŧ	0.029	9.61×10^{-7}	-	I	1	$PDE9A^{\dagger}$
	2	126264896	126316073	9	rs482380239 ^b	126278844	υ	0.989	4.09×10^{-7}	+	I	1	FGR^{\dagger}
	4	10350276	10372066	ŝ	rs523829414°	10369692	A	0.005	5.84×10^{-7}	+	I	1	HEPACAM2 [†]
	5	105753443	105755633	2	rs209426521ª	105753443	F	0.015	9.60×10^{-7}	+	+	0	N/A
	9	38518710	39614235	4	$rs111701266^{a}$	39097535	Ċ	0.007	1.71×10^{-7}		+	7	NCAPG,
													LCORL
	9	91602424	91626735	7	$rs210873983^{b}$	91602424	A	0.011	1.76×10^{-8}	++	++++	1	$PARM1^{\dagger}$
	7	37616120	37616155	2	$rs378671192^{a}$	37616155	A	0.006	1.71×10^{-7}	++		0	N/A
	10	72286947	72287008	ŝ	$rs381793861^{a}$	72286947	Ŧ	0.009	5.59×10^{-7}	I	++++	0	N/A
	12	14636353	14960925	ŝ	$rs800390196^{a}$	14750727	A	0.014	2.63×10^{-7}	+	+	2	TSC22D1
	12	56675034	56682826	Ŋ	rs209495875ª	56676660	A	0.044	1.04×10^{-8}	+	I	0	N/A
	24	19405971	19427409	S	$rs109333678^{a}$	19405971	ტ	0.995	8.17×10^{-8}	+++		0	N/A
CH	2	29919882	29920135	ŝ	$rs380146645^{a}$	29919882	Ċ	0.017	8.67×10^{-7}		I	0	N/A
	11	92402757	92405438	12	rs445896050ª	92402757	Η	0.006	2.26×10^{-9}		+	0	N/A
	20	44514471	44520630	9	$rs379731054^{a}$	44518024	A	0.014	2.35×10^{-7}			0	N/A
HF	26	1759408	2501334	ŝ	rs133176675ª	1765944	A	0.924	8.48×10^{-8}	++		0	N/A
	26	49916091	49918252	2	rs110755029ª	49918252	υ	0.788	1.83×10^{-7}	++	+	0	GLRX3
	26	51542180	51546414	ŝ	$rs209816010^{b}$	51542180	Η	0.892	6.16×10^{-7}	+++	+	1	STK32C [†]
ILM	7	40539352	40546791	2	rs380343675ª	40539352	υ	0.032	7.11×10^{-7}	1	+	0	EPHA6
¹ No. SNP is tl difficulty. SN size for DCD the SNP was present in th identified in	is the num SNP effect CD, i.e., a S vasn't pres 1 the QTL t in.	ber of SNP withi DCD is the QTL SNP effect size in ent in that datas he nearest funct	¹ No. SNP is the number of SNP within the QTL with a P-value $\leq 1 \times$ difficulty. SNP effect DCD is the QTL-lead SNP's direction effect for size for DCD, i.e., a SNP effect size in the top 20%, – represents opp the SNP wasn't present in that dataset due to allele frequency. SNI present in the QTL the nearest functional candidate gene within 2 identified in.	P-value - tion effec epresents requency gene with	\$1 × 10 ⁻⁶ . P-value is it for direct calving s opposing SNP effe- : SNP effect PM sim in 250 kb was chos	10 ⁻⁶ . <i>P</i> -value is the <i>P</i> -value of strongest SNP association. + Allele is the allele that had a positive SNP effect on maternal calving c direct calving difficulty, + represents SNP effects in the same direction, ++ represents the same direction and large SNP effect ossing SNP effect and <i>P</i> effect directions that have a large SNP effect size. <i>N</i> /A (Non-applicable) was if <i>P</i> effect PM similar to DCD but represents the SNP effects for PM. No. of genes is the number of genes in the QTL. If no gene was is 50 kb was chosen. ^a intergenic, ^b intronic, ^c downstream gene variant, [†] gene in which the most significant SNP within the QTL was	agest SNP asso nts SNP effects ts opposing SN resents the SNF ronic, ^c downstr	ciation. + All i in the same P effect dire effects for eam gene v	ele is the allele t e direction, ++ rel ctions that have PM. No. of genes ariant, †gene in w	hat had a posit presents the sa a large SNP eff is the number of hich the most	ive SNP effec me direction ect size. N/A of genes in th significant SI	:t on mater and large (Non-appli ne QTL. If <i>n</i> NP within t	nal calving SNP effect cable) was if o gene was he QTL was

Table 3. QTL associated with maternal calving difficulty 1

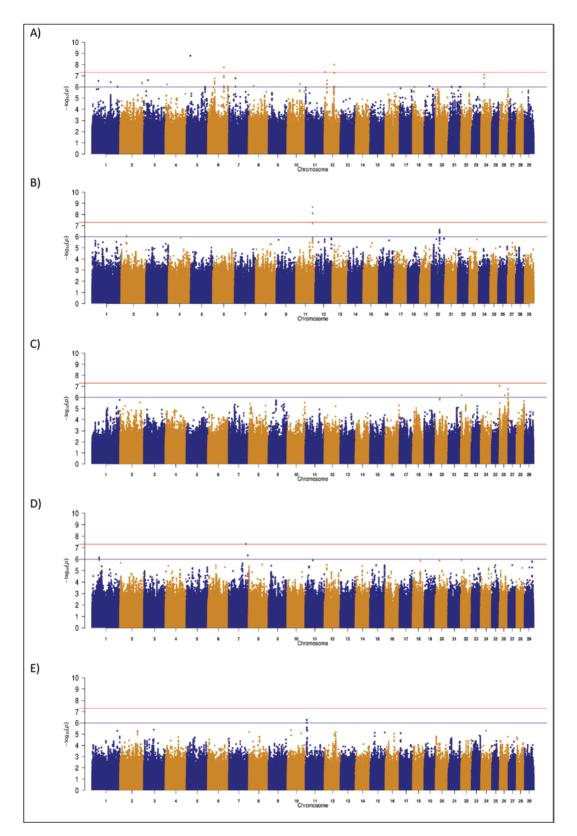


Figure 3. Manhattan plots for maternal calving difficulty in (A) Angus, (B) Charolais, (C) Holstein Friesian, (D) Limousin, and (E) across all breeds. The red line indicates the genome-wide significance threshold of $P \le 5 \times 10^{-8}$ and the blue is the suggestive significance threshold of $P \le 1 \times 10^{-6}$.

SNP with PM adjusted for DCD with sodium voltage-gated channel beta subunit 4 (SCN4B) being the likely putative candidate gene; the minor allele (G) of the strongest SNP

in this QTL was associated with a reduction in PM and had a moderate allele frequency of 0.153. No missense variant within SCN4B was segregating within the CH population

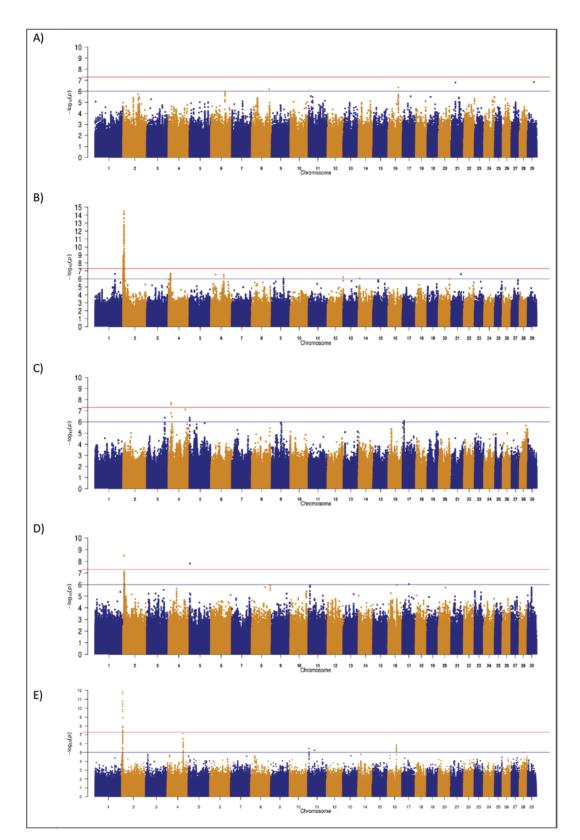


Figure 4. Manhattan plots for direct mortality in (A) Angus, (B) Charolais, (C) Holstein Friesian, (D) Limousin, and (E) across all breeds. The red line indicates the genome-wide significance threshold of $P \le 5 \times 10^{-8}$ and the blue is the suggestive significance threshold of $P \le 1 \times 10^{-6}$.

in the current data set although several intronic variants within SCN4B did have P-values ranging from 5.66 \times 10⁻⁶ to 9.75 \times 10⁻⁵.

Pleiotropic associations

The number of SNP associated (i.e., P-value $<10^{-4})$ with more than one calving trait differed by breed; no SNP was associated

				No.		Strongest		+ Allele		SNP effect	SNP effect	No.	
Breed	BTA	Start	End	SNP	Strongest SNP	SNP position	+ Allele	freq	P-value	DCD	MCD	genes	Candidate gene(s)
AA	∞	100644490	100645055	2	rs797916567 ^b	100644490	A	0.016	6.69×10^{-7}	+	I	1	PTPN3 ⁺
	16	58808269	58818335	2	$rs134283619^b$	58814914	A	0.019	4.38×10^{-7}	+		1	$RFWD2^{\dagger}$
CH	2	301926	957129	18	$rs43598129^{a}$	900492	ტ	0.050	1.42×10^{-7}	‡	I	ß	HERC2, NIPA1, OCA2
	2	1007617	1226503	5	$rs458371570^{a}$	1110887	A	0.039	3.79×10^{-9}	‡	I	2	CYFIP1, TUBGCP5
	2	1296023	3143209	125	$rs110820498^{a}$	1925667	H	060.0	1.31×10^{-9}	‡	I	4	N/A
	2	3278825	3626278	28	$rs461349284^{a}$	3626278	A	0.051	4.89×10^{-8}	‡	I	0	N/A
	2	4580833	4602350	7	rs383237760ª	4602350	A	0.774	4.46×10^{-7}	‡	+	0	N/A
	2	5305742	6938016	654	rs799943285ª	6808074	ტ	0.056	6.56×10^{-15}	‡		17	MSTN
	2	7334284	7609482	S	$rs798834436^{a}$	7488750	Г	0.052	1.54×10^{-7}	‡		1	COL3A1
	2	7773097	7776214	ę	$rs210662731^{a}$	7773097	H	0.369	3.39×10^{-7}	‡	I	0	GULP1
	4	12723521	12787299	139	$rs381171531^{a}$	12733539	υ	0.040	2.09×10^{-7}	I	I	2	PDK4
	9	73569965	73700948	2	rs43473352°	73636267	ტ	0.935	2.93×10^{-7}	+	+	2	HOPX
	6	68152800	68267416	∞	rs379327303ª	68152800	A	0.988	8.85×10^{-7}	1	+	1	LAMA2
	14	8591593	8591601	2	rs526183014	8591601	A	0.552	8.44×10^{-7}	+	+	0	ZFAT
HF	ŝ	101909079	102183746	11	$rs378988599^{b}$	102122935	H	0.818	4.30×10^{-7}	+	I	ŝ	C1 orf228, TMEM53, RNF220 [†]
	4	16001571	16013815	4	$rs209481751^{a}$	16013815	A	0.539	1.87×10^{-8}	+	+	0	N/A
	Ŋ	1492436	1503881	4	$rs438879818^b$	1503881	H	0.086	4.11×10^{-7}	I	+	1	$TBC1D15^{\dagger}$
	17	8589800	8600386	2	rs382728302ª	8600386	A	0.040	8.11×10^{-7}	+	I	0	N/A
LM	2	6481284	6751879	47	$rs110953984^{a}$	6655139	A	0.031	7.76×10^{-8}	+		9	MSTN
	2	7503983	7562999	21	$rs210765106^{a}$	7505936	A	0.103	1.19×10^{-7}	‡		0	N/A
	2	7625971	7819581	11	$rs378228653^{b}$	7787549	H	0.931	2.40×10^{-7}	+		2	$GULP1^{\dagger}$
	2	8147499	8314779	14	rs800119023ª	8268605	A	0.010	3.63×10^{-7}	+		0	N/A
	2	9497001	9502488	4	$rs382982814^{a}$	9497001	Т	0.057	1.38×10^{-7}	+		2	ZSWIM2,
	Ŋ	2155306	2167134	2	$rs378053936^{b}$	2167134	υ	0.118	1.56×10^{-8}	+	I	2	TRHDE ⁺
¹ No. SN difficult	P is the ni y. SNP eff	umber of SNP w	'No. SNP is the number of SNP within the QTL with a P-value \le 1 × difficulty. SNP effect DCD is the QTL-lead SNP's direction effect for	rith a P-v direction	¹ No. SNP is the number of SNP within the QTL with a P-value $\leq 1 \times 10^{-6}$. P-value is the P-vidifficulty. SNP effect DCD is the QTL-lead SNP's direction effect for direct calving difficult	ue is the P-value o /ing difficulty; + re	f strongest : presents SN	SNP associat P effects in 1	ion. + Allele is th the same direction	ie allele tha on, ++ repre	t had a pos esents the s	sitive SNP e same direc	10 ⁻⁶ . P-value is the P-value of strongest SNP association. + Allele is the allele that had a positive SNP effect on maternal calving direct calving difficulty; + represents SNP effects in the same direction, ++ represents the same direction and large SNP effect

size for DCD, i.e., a SNP effect size in the top 20%, – represents opposing SNP effect and – represents opposing SNP effect directions that have a large SNP effect size. N/A (Non-applicable) was if the SNP wasn't present in that dataset due to allele frequency. SNP effect MCD is similar to DCD but represents the SNP effects for maternal calving difficulty. No. of genes is the number of genes in the QTL. If no gene was present in the QTL the nearest functional candidate gene within 250 kb was chosen. ^aintergenic, ^bintronic, ^cdownstream gene variant, ¹gene in which the most significant SNP within the QTL was identified in.

Table 4. QTL associated with direct PM^1

with more than one trait in the HF population, 5 SNP were pleiotropic in the AA population, 116 were pleiotropic in the LM population, and 882 SNP were pleiotropic in the CH population (Supplementary Figure 4). Of the 882 SNP identified in the CH population that were associated with both DCD and PM, all but one were located on BTA2 residing between 1.22 and 8.07 Mb and all had the same SNP effect direction for both CD and PM. This included the Q204X mutation where the minor T allele was associated with an increase (i.e., worsening) in both DCD and PM but was not associated (P = 0.11) with MCD. The remaining SNP identified in the CH population was rs384777136, an intergenic variant on BTA10 whose minor allele (MAF = 0.012) increased both DCD (P = 2.35 \times 10⁻⁵) and PM (P = 4.87 \times 10⁻⁵). Similar to the CH population, the overlap between DCD and PM in the LM population was primarily on BTA2 spanning from 5.94 to 9.49 Mb with 111 SNP in this region having the same effect direction for both traits and a P-value < 10⁻⁴. Only two SNP, both located on BTA2, were pleiotropic with a P-value $< 10^{-4}$ across all the three calving traits in the LM population; the minor alleles at both SNP were associated with a deterioration in DCD and PM but an improvement in MCD. The minor allele at SNP rs799053464 on BTA5 which had a P-value < 10^{-4} for both DCD and MCD in the AA population also differed in SNP effect direction for both traits. Overlap between DCD and PM was also detected in the AA population on BTA8 in the 13.45 to 13.47 Mb region where four intergenic SNP had P-values ranging from 6.31×10^{-5} to 9.94×10^{-5} for DCD and from 2.78×10^{-5} to 4.21×10^{-5} for PM; the minor allele for all four SNP, which had a mean allele frequency of 0.015, was associated with a decrease (i.e., better) in EBV for all three calving traits.

More overlapping genomic regions among all three calving traits were detected when 500-kb windows rather than individual SNP were considered to identify pleiotropic genomic regions. Genomic regions that contained a SNP with a *P*-value < 10^{-4} associated with all calving performance traits existed in each breed, although the degree of overlap differed by breed (Figure 5); overlap ranged from one 500-kb window in the HF to seven regions in each of the AA and CH populations. The seven

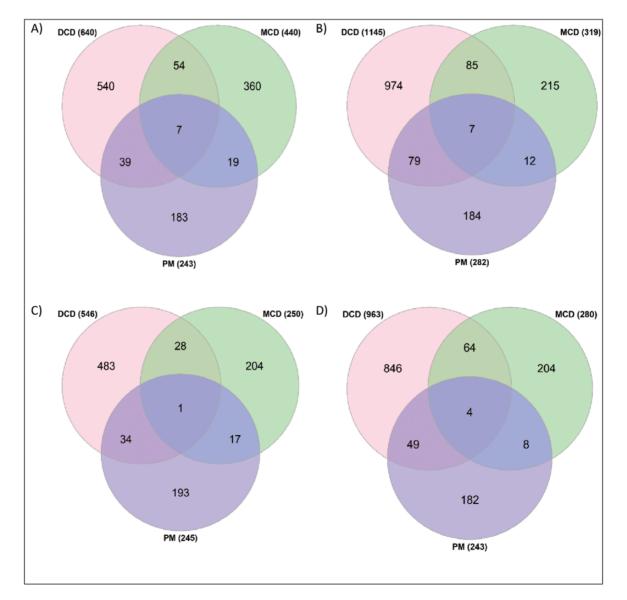


Figure 5. The number of 500-kb windows containing an associated SNP with a P-value < 10⁻⁴ that overlapped across traits within each breed. Each window had to contain a SNP with an unadjusted P-value < 10⁻⁴. (A) Angus, (B) Charolais, (C) Holstein-Friesian, and (D) Limousin. Pink represents direct calving difficulty (DCD), green represents maternal calving difficulty (MCD), and blue represents PM.

genomic windows in the AA population were located across five chromosomes (BTA6, 7, 13, 22, and 24) with multiple underlying plausible candidate genes including HPGDS (BTA6), RBMS3 (BTA22), NCPC, and MYO5B (BTA24). The MCD-associated SNP in each window was in antagonistic pleiotrophy with either DCD or PM in six of the seven windows. Five of the seven windows identified in the CH population to be associated with all three calving traits were located in the first 4.5 Mb of BTA2, whereas the remaining two windows were located on BTA12 and BTA22 encompassing GPC6 and LZTFL1, respectively. The only window identified in the HF to be associated with all three calving traits was located on BTA7 where proline rich 16 (PRR16) was a plausible candidate gene; each SNP with a P-value < 10⁻⁴ within the window for each of the calving traits had the same SNP effect direction for all three traits.

A greater percentage of window overlap was evident between DCD and PM rather than with MCD; for example, in the CH, 46.74% of the PM-associated windows were also associated with DCD while only 10.37% were associated with MCD. Despite detecting window overlap across all traits, the majority of genomic windows were trait-specific. As the majority of SNP were trait-specific, the allele substitution effects with a P-value < 10^{-4} for an individual calving trait were investigated across each of the traits (Supplementary Figure 5). A greater percentage of SNP with a P-value < 10^{-4} for DCD had opposing SNP substitution effects with MCD than those with PM. For example, only 31.04% of the SNP with a P-value < 10^{-4} for DCD in the CH population had allele substitution effects which were opposite in sign for PM, whereas 59.88% of SNP had opposing allele substitution effects for MCD.

Discussion

This association study relating the bovine genome to calving performance, which is one of the largest studies to date, identified multiple underlying QTL both within and across dairy and beef breeds; the majority of the associations, however, appear to be both breed- and trait-specific. The existence of breed-specific SNP effects and trait-specific SNP effects for calving performance traits has been previously documented (Purfield et al., 2015; Tiezzi et al., 2018); thus, highlighting the complexity of the genomic features influencing the expression of these traits. The high proportion of trait-specific QTL identified in the present study was unexpected given the existence of a moderate negative genetic correlation between the direct and maternal components of calving difficulty (Crowley et al., 2011); this correlation manifests itself as a small easy calving female calf, once mature, having a greater likelihood of experiencing a difficult calving due to her reduced size and pelvic opening dimensions (Philipsson, 1979). Therefore, genomic regions common to both DCD and MCD were expected. Understanding the genomic architecture influencing the direct and maternal components of calving difficulty separately and the known genetic antagonism (Crowley et al., 2011) may aid in effective simultaneous selection for both traits as the genetic linkage across traits could be potentially resolved during meiosis once the underlying causal mutations are known. However, additional analyses with the new bovine reference genome ARS-UCD 1.2 and the ever-increasing quantity of gene expression analyses would be required to functionally characterize any putative candidate gene identified in the present study.

Antagonistic pleiotropy

A limited number of pleiotropic SNP were identified in the beef populations, although they were primarily located on BTA2 either within or near to the MSTN gene; these SNP were, however, only associated with DCD and PM and not with MCD. Heavier cattle birth weight has been previously reported for double-muscle carriers which ultimately lead to an increase in the incidence of calving difficulty and PM within South Devon cattle (Wiener et al., 2002).

In contrast to previous calving association studies (Thomasen et al., 2008; Sahana et al., 2011), no pleiotropic SNP was identified in the HF population despite the known strong genetic correlation between DCD and PM (Steinbock et al., 2003) and the moderate negative correlation between DCD and MCD (Eaglen and Bijma, 2009). The inability to identify pleiotropic calving SNP is most likely a reflection of the polygenic architecture in the HF, which is also reflected in the low heritability of MCD and PM in HF populations (Eaglen and Bijma, 2009) in comparison to beef populations (Mujibi and Crews, 2009). In addition, difficult calving bulls are simply not tolerated within the dairy industry due to the adverse effects of a difficulty calving on both milk performance and subsequent fertility (Dematawewa and Berger, 1997; Berry et al., 2007; Eaglen et al., 2011). Therefore, only a limited number of SNP of large effect was associated with calving performance in the HF population in comparison to the beef populations in the present study. The smaller SNP effect size and the limited role of major genes in calving performance in the HF population may be why a greater percentage of SNP had the same SNP effect direction for DCD and MCD than the beef populations; in total 83.01% of the 7,533 SNP identified in the HF population (Supplementary Figure 4) had the same SNP effect direction for both DCD and MCD, whereas in comparison 37.97% and 31.72% of SNP associated with DCD in the CH and LM SNP shared effect directions with MCD.

As the characteristics of a difficult calving are associated with an increased risk of PM (Mee et al., 2008) a greater proportion of SNP sharing the same SNP effect direction is likely and this was corroborated in the beef populations used in the present study. The use of 500-kb windows identified additional putative pleiotropic genomic regions associated with calving performance within each of the breeds but the antagonistic pleiotropy that existed between DCD and MCD remained in the beef populations. Although antagonistic pleiotropy was detected between DCD and MCD, QTL associated with DCD where the lead SNP had a small opposing SNP effect on MCD compared to that for DCD were identified (Table 1) and vice versa (Table 2). This suggests that genetic gain for improved calving performance could be achieved by identifying trait-specific QTL of large effect that have a minimal effect on their correlated trait.

The role of BTA18 in calving difficulty

Chromosome 18 has long been associated with calving difficulty in multiple Holstein populations (Cole et al., 2009, 2011; Sahana et al., 2011; Höglund et al., 2012; Purfield et al., 2014, 2015) suggesting that a true QTL association impacting calving ability exists on this autosome. Indeed the same genomic region has also been associated with multiple fertility-related traits and body conformation traits (Abo-Ismail et al., 2017; Muller et al., 2017; Doyle et al., 2020a). Despite several studies proposing the same genomic region on BTA18, consistency is lacking on the suggested underlying candidate gene. Cole et al. (2009) was the first to propose that the pleiotropic effects of this QTL may be due to the QTL involvement in embryo development and prolonged gestation thus increasing calving difficulty, and subsequently impacting fertility performance. This theory was further substantiated by recent research from Fang et al. (2019) who proposed ZNF613 as a plausible candidate gene within this region that was likely causing a lengthening of gestation within a large U.S. Holstein population of 27,214 Holstein bulls. Fang et al.

(2019) demonstrated strong evidence that it was the methylation pattern of the second intron of ZNF613 that was lengthening gestation and found this pattern to be also associated with sire calving ease, body depth, and conception rate in Holsteins. The downstream variant of ZNF613, rs381577268, which was the strongest association with DCD in our HF population, was also the strongest association for gestation length in a HF population overlapping in animals used in the present study (Purfield et al., 2019b). Animals homozygous for the T allele at this position in the present study had a 4.7-d longer gestation length and, on average, a worse (P < 0.001) EBVs for DCD and PM (P < 0.05) than homozygous noncarriers (Purfield et al., 2019b). Despite the strong evidence corroborating ZNF613 as the plausible candidate gene, limited genetic gain in calving difficulty is likely to be achieved through its targeted selection in a breeding program as the frequency of the favorable (i.e., C) allele was near-fixation in both Irish (0.98) and U.S. Holstein populations (0.93). This QTL on BTA18 is also likely to have a limited role in improving across-breed genomic evaluations, as several studies involving other dairy breeds such as Jersey (Tiezzi et al., 2018), Fleckvieh (Pausch et al., 2011), and Brown Swiss (Frischknecht et al., 2017; Tiezzi et al., 2018) failed to associate this QTL with calving performance. In addition, no SNP within ZNF613 was suggestively associated with DCD in the any of the beef populations investigated in the present study. However, this may be due to a difference in the linkage phase between the imputed alleles and causal mutations in each of the breeds, as the marginal significance of SNP ($P > 1 \times 10^{-6}$) just upstream of this QTL in the CH and LM populations does suggest that a QTL extending from the 57.9 to 58.2 Mb on BTA18 influences DCD in multiple breeds, albeit to a different extent.

Additionally known calving performance QTL

The role of MSTN in calving performance has long since been established (Casas et al., 1999; Bellinge et al., 2005). In the present study, QTL regions containing MSTN were associated with both DCD and PM in the CH population, whereas in the LM population, the QTL associated with these traits were located just downstream of MSTN. The well-known stop-gain Q204X mutation within the CH population contributed to 5.09% of the genetic variance in DCD. Interestingly the F94L mutation in MSTN, which is known to result in a moderate muscle hypertrophy phenotype in the LM population, was not associated with any calving performance trait in the present study. This is not unexpected as previous studies have failed to detect a difference in birth weight in LM between F94L carriers vs. noncarriers (Esmailizadeh et al., 2008). This is also further substantiated by the lack of an association of the F94L mutation with DCD within the CH population; in total 282 homozygous F94L carriers existed within the CH population and the P-value for F94L was 0.189. The nonsignificant association of F94L with calving performance suggests that the targeted selection of this mutation for increased carcass performance (Esmailizadeh et al., 2008) within both the CH and LM populations can be achieved without the adverse effects of greater dystocia; the same, however, is not true for Q204X.

The known strong genetic correlation between calving difficulty and both birth weight and body size (Johanson and Berger, 2003) implies that genomic regions and candidate genes previously associated with cattle height should possibly be in common with those associated with calving performance in the present study. The PLAG1 gene, which has been previously associated with height in both humans (Gudbjartsson et al., 2008) and cattle (Pryce et al., 2011; Bouwman et al., 2018), as well

as live weight in cattle (Littlejohn et al., 2012; Nishimura et al., 2012), was indeed associated with DCD in the HF population used in the present study. Although PLAG1 has been associated with calving performance in multiple breeds, such as Fleckvieh (Pausch et al., 2011) as well as U.S. Gelbvieh and Simmental (Saatchi et al., 2014), this QTL region on BTA14 was only associated with DCD in the HF population in the present study despite the segregation of variants within 250 kb up/downstream of PLAG1 in each of the beef populations; this included the seven PLAG1 variants identified by Karim et al. (2011) which had MAFs ranging from 0.014 to 0.024 in the AA population, from 0.011 to 0.013 in the CH population, and from 0.032 to 0.035 in the LM population. Interestingly, a QTL on BTA6, which contained the stature-related genes, NCAPG and LCORL, was also associated with MCD but only within the AA population. This region has been previously associated with calving performance in several populations (Olsen et al., 2010; Bongiorni et al., 2012; Saatchi et al., 2014) and was the strongest QTL association for height of withers within an Irish AA population (Doyle et al., 2020b). The minor alleles of the four associated SNP ($P < 1 \times 10^{-6}$) identified within this QTL (Table 3) were below the MAF threshold for inclusion in the present study in the remaining breeds suggesting this QTL is of greater importance in the AA population which is characterized as being phenotypically smaller than the CH, HF, and LM populations (Bouwman et al., 2018).

Novel associations

Although several major well-known genes attributed to calving performance were confirmed in the present study, several novel candidate genes are also proposed. Interestingly, PDK4 which was suggestively associated with PM in the CH population, and identified as a putative across-breed association in the window analyses in the CH, HF, and LM populations, has been documented to play a role in the glucose homeostasis during the embryo development in both pigs (Lan et al., 2009) and cattle (Lonergan et al., 2016). A 16-bp deletion within PDK4 has also been attributed to dilated cardiomyopathy within the Doberman Pinscher canine breed, which ultimately results in their sudden death as their hearts are unable to generate sufficient energy for long-term healthy function (Bolfer et al., 2016). The allele frequency of the favorable allele of the strongest SNP in this QTL (rs381171531), which was associated with a reduction in PM, was highly prevalent (0.959) in the CH population; nonetheless, some genetic gain could still be made through its targeted selection while having limited impact on both DCD and MCD.

The strongest association in the AA population for DCD encompassed the GC gene which encodes the vitamin D-binding protein. Although the calf itself cannot generate vitamin D during gestation, it is suggested that this QTL is impacting DCD by affecting the calf's birth weight. Studies in humans have associated several GC variants with a lighter birth weight in infants due to a reduction in circulating vitamin D levels (Chun et al., 2017). Interestingly, this region has been previously associated with gestation length in Holstein-Friesians (Maltecca et al., 2009) and GC was identified as a putative candidate gene for suckling traits in beef cattle (Michenet et al., 2016). The association of variants downstream of the GC gene suggests that it is the regulatory expression of GC that is influencing DCD rather than a disruptive mutation. Another putative candidate gene identified in the AA population was PARM1 which contained several intronic variants that were suggestively associated with MCD. PARM1 has been previously shown to play a role in the development of the blastocyst during early embryonic development (Zolini et al., 2019) but it is unclear

what role it could have on MCD. The across-breed analyses of 47,753 sires also identified another putative candidate gene associated with MCD on BTA11; SLC9A2 has been shown to be differentially regulated over the course of gestation in humans and may play a role in intrauterine growth restriction (Johansson et al., 2002). The candidate gene, ZEB2 which was identified as a candidate gene for DCD in the AA, LM, and HF populations from the window analyses has been previously associated with birth weight in Brahman cattle (Martinez et al., 2017) and its deletion has also been documented to cause polled and multisystemic syndrome in CH cattle (Capitan et al., 2012).

Novel candidate genes associated with all three calving traits included PRR16 which was identified in the HF population. Largen, the product of PRR16, has been shown to be essential in maintaining cell size homeostasis with overexpression of this gene in transgenic mice resulted in embryonic death (Yamamoto et al., 2014). PRR16 is highly expressed in both human placental endothelial cells and also in the bovine endometrium. Although it has been shown to regulate cell size, PRR16's role in calving performance is unclear but may affect performance through the regulation of mitochondrial functions and mRNA translation.

Across-breed associations

Despite the majority of detected associations being traitspecific, common genomic regions that were associated with calving performance across the different breeds were identified and could help in possibly facilitating across-breed genomic predictions. The greatest percentage of overlap was for DCD between the CH and LM populations with 311 SNP surrounding MSTN having the same sign of allele substitution effect for DCD. Saatchi et al. (2014) previously reported that major genes such as MSTN were pleiotropic for DCD in multiple cattle breeds and that these QTL also explain the majority of genetic variance in birth, weanling, yearling, and carcass weight. Aside from wellknown major genes, multiple novel genes such as ZEB2 (DCD), SLC9A2 (MCD), and PDK4 (PM) were also identified, but the proportion of genetic variance explained by these candidates was minimal thus reaffirming that the majority of variance in calving performance is attributed to the additive (and possibly multiplicative) effect of many polymorphisms of small effect. Therefore, to accommodate breed-specific major genes of large effect, it is suggested that genomic prediction algorithms should be potentially partitioned into a within-breed component and an across-breed component for calving performance.

Mortality adjusted for calving difficulty

The association between dystocia and PM in cattle is wellknown (Mee et al., 2008). Little, however, is known about the prevalence of PM in unassisted calving events. There are many possible causes for such mortality events including congenetial defects, placental dysfunction (Berglund et al., 2003), infections (Givens and Marley, 2008), precalving nutrition (Mee et al., 2013), or possibly management issues such as age at first calving (Mee et al., 2008). Indeed several lethal recessive mutations and haplotypes have been identified in cattle in recent years (VanRaden et al., 2011; Fritz et al., 2013; Daetwyler et al., 2014; Adams et al., 2016). Three well-known attributers to calf mortality include complex vertebral malformation, bovine leukocyte adhesion deficiency, and brachyspina, and although the causative mutation within each gene is known, none of these mutations were included in the present data set and no segregating variant within their causative genes had a P-value < 10^{-4} for PM. Interestingly, the causative mutations identified in the HF lethal haplotypes HH1 (rs448942533) and HH3 (rs456206907) were segregating within our HF population but were not significant for PM with P-values of 0.83 and 0.54, respectively.

Several QTL regions in the present study were associated with mortality adjusted for genetic merit for DCD in the CH population but only two QTL had not been previously identified in the PM analysis. The continued association of QTL surrounding MSTN and PDK4 suggests that these candidate genes remain associated with PM even when calving difficulty is accounted for. SCN4B, which was identified as a novel putative candidate gene for PM not associated with dystocia, has been previously associated with atrial fibrillation in humans (Li et al., 2013). Although further investigation into the putative role of SCN4B in PM within the CH population is needed, the moderate allele frequency of the G allele (MAF = 0.153) for the strongest SNP association within this QTL suggests genetic improvement could be made if selected upon.

Designer matings using genomic information

Prediction of genetic merit for calving performance traits using genomic selection methodology, like most traits, is generally undertaken at a macro level by summing the genomic feature effects across the entire genome (Hayes et al., 2009). No distinction is generally made between the biological basis as to why a particular parent may generate calves with a greater than average risk of either calving difficulty or mortality, and how that biology complements the genomic features of its mate. Feto-pelvic disproportion is one of the most common risk factors associated with dystocia in cattle (Mee et al., 2008). Hence, mating of individuals compatible on feto-pelvic dimensions based on specific genomic predictions could help advert the risk of calving dystocia and, by association (Berry et al., 2007), calf mortality. One clear example of the potential of exploiting underlying genomic information is that related to mutations within the MSTN gene and risk of dystocia. MSTN, which is also known as growth differentiation factor 8 (GDF-8), is expressed in developing and mature skeletal muscle (McPherron and Lee, 1997); MSTN activity represses skeletal growth. Therefore, animals, especially with the mutant nt821 variant, exhibit muscular hypertrophy (Grobet et al., 1997). Wiener et al. (2002) reported a greater calving difficulty in cattle carrying the nt821 (they called this del11) mutation. While the nt821 mutation in the MSTN gene has a clear major gene effect, one of the motivations in the present study was to identify other potential variants associated with animal conformation which, when summed together, could provide a more in-depth assessment of the likely contributing factors to eventual dystocia. Upperman et al. (2019) proposed a strategy for recommending matings among beef animals carrying lethal mutations; a similar strategy could be deployed for calving performance but instead of the characteristic of interest being a lethal mutation, it could be genetic predisposition to dystocia or PM. Mates both with a high genetic risk, and underpinned by the same biological rational, could then be avoided. Other than MSTN, several other genes such as PLAG1, GC, and NCAPG/LCORL were associated with body size and such knowledge could help assess the likelihood of dystocia in a future mating given the genotypes of both candidate mates.

Being able to disentangle to genetic architecture of calving performance traits, as well as being useful in mating programs, can also be useful in breeding schemes. Many traits in breeding goals where selection in the same direction is favored can be antagonistic. The objective of an efficient breeding program is to identify animals that excel in both traits, despite the antagonistic correlation. Genetic correlations are due to either pleiotropy or linkage. Falconer and Mackay (1996) suggested that simultaneous selection for two traits in the same direction will, over time, cause an antagonistic genetic correlation to develop and this will be due to pleiotropy. While sometimes disputed (Robinson, 1996), most studies in cattle (Crowley et al., 2011) have reported an antagonistic genetic correlation between DCD and MCD. The favored direction of selection for both traits is the same. Hence, of interest is the ability to identify genomic regions with allele substitution effects for both traits in the same direction, or with an estimated effect for one of the traits but not for the other. The strongest SNP within several QTL in the present study (Tables 1-3) had the same SNP effect direction across all calving traits and of those QTL that had opposing effects, the SNP effect was small for the other calving traits.

Conclusions

Several genomic regions associated with calving performance in dairy and beef cattle were identified in the present study although the majority of these were unique to a given breed or trait. Major genes previously documented to be associated with calf body size such as MSTN, PLAG1, and NCAPG/LCORL were associated with calving performance and should be exploited within genomic mating programs to predict the likelihood of a dystocia event, and, where necessary minimize the risk. Novel putative candidates such as PARM1, PDK4, and SLC9A2 were also identified, although the proportion of genetic variance explained by these candidates was small. This substantiates the theory that calving performance traits are complex quantitative traits under the influence of many polymorphisms each of small effect. Antagonistic pleiotropy between DCD and MCD was detected, although several QTL regions that had the same SNP effect direction across both traits were identified suggesting simultaneous genetic improvement for both traits through targeted selection is feasible once the genomic architecture is known.

Supplementary Data

Supplementary data are available at Journal of Animal Science online

Acknowledgments

Funding from the European Union's Horizon 2020 research and innovation program – GenTORE – under grant agreement no. 727213 is greatly appreciated, as is the funding from the Science Foundation Ireland principal investigator award grant (14/IA/2576) and Research Centers (16/RC/3835; VistaMilk). We gratefully acknowledge the 1000 Bull Genomes Consortium for providing accessibility to whole-genome sequence data which was used in this study and to the Irish Cattle Breeding Federation for providing access to genotype and phenotype data.

Conflict of interest statement

The authors declare no real or perceived conflicts of interest.

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