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### Genetic determinants for intramuscular fat content and waterholding capacity in mice selected for high muscle mass

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#### Abstract

Intramuscular fat content and water-holding capacity are important traits in livestock as they influence meat quality, nutritive value of the muscle, and animal health. As a model for livestock, two inbred lines of the Berlin Muscle Mouse population, which had been long-term selected for high muscle mass, were used to identify genomic regions affecting intramuscular fat content and water-holding capacity. The intramuscular fat content of the *Musculus longissimus* was on average 1.4 times higher in BMMI806 than in BMMI816 mice. This was accompanied by a 1.5 times lower water-holding capacity of the *Musculus quadriceps* in BMMI816 mice. Linkage analyses with 332 G<sub>3</sub> animals of reciprocal crosses between these two lines revealed quantitative trait loci for intramuscular fat content on chromosome 7 and for water-holding capacity on chromosome 2. In part, the identified loci coincide with syntenic regions in pigs in which genetic effects for the same traits were found. Therefore, these muscle-weight-selected mouse lines and the produced

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intercross populations are valuable genetic resources to identify genes that could also contribute to meat quality in other species.

#### Introduction

Meat is a major protein source in human nutrition and the consumption of meat products is growing worldwide. Understanding the genetic and metabolic backgrounds of muscle tissue characteristics is important for the sustainable improvement of meat quality but also for human and livestock health. The aim of this study was to localize regions in the mouse genome that regulate muscle traits that are important for meat production in livestock. The results contribute to improving the knowledge of genomic regions that are involved in economically important traits in animal production.

Intramuscular fat content (IMF) affects glucose uptake and the metabolism and health of humans and animals. Elevated levels of intramyocellular lipids can be a sign of impaired utilization of fatty acids as energy fuel or even insulin resistance (Phillips et al. 1996; Yaspelkis et al. 2004). Besides the health aspect, IMF greatly impacts the taste and tenderness of meat products (Cannata et al. 2010; Fernandez et al. 1999). In farm animal species, the amount of IMF and its distribution in the muscle differ between breeds, which suggests genetic components for this trait.

So far, only a few quantitative trait loci (QTL) for IMF have been mapped in sheep, pigs, and cattle (de Koning et al. 1999; Lambe et al. 2010; Underwood et al. 2007), and some candidate genes, e.g., porcine delta-like 1 homolog (*DLK1*), fatty acid binding protein 3 (*FABP3*), and the leptin receptor (*LEPR*), have shown associations with that trait (Schwab et al. 2009; Tyra et al. 2010; Zhao et al. 2010). In mice, for example, *Ccr2* was described as an IMF-influencing gene in a knockout model (Ochoa et al. 2006).

Genetic determinants that have an impact on postmortem muscle traits are hardly known. This is especially the case for the water-holding capacity (WHC) of meat, which is the ability of the muscle to store tissue fluids within its fibers after the cessation of normal physiological conditions. Low WHC leads to impaired meat-processing properties and does not meet consumers' preferences. Several QTL and candidate genes for WHC, e.g., ryanodine receptor 2 (*RYR2*), integrins,  $\mu$ -calpain, and desmin, have been found in pigs (Bee et al. 2007; Fujii et al. 1991; Malek et al. 2001; Ponsuksili et al. 2008; Wimmers et al. 2010; Zhang et al. 2006).

Because the extent and influence of genetic regulators on IMF and WHC are not yet fully understood, we performed a crossbred experiment between two phenotypically different Berlin Muscle Mouse inbred lines that have been long-term selected for high muscle mass before inbreeding. The study aimed to identify QTL that affect IMF and WHC. In addition, we analyzed traits like muscle glycogen and lactate contents, blood glucose levels, and body composition which might correlate with IMF and WHC.

#### Materials and methods

#### Animals

The Berlin Muscle Mouse population had been long-term selected for high body weight and high muscle mass to reflect the selective mechanisms in livestock breeding. Founder animals of the Berlin Muscle Mouse (BMM) population were originally purchased in several pet shops in Berlin, Germany. The selection process comprised several distinct phases. The beginning of the selection process constituted a phase of 23 generations of selection for high

protein content of the carcass at the age of 60 days. Protein content was determined by chemical analyses. In a second phase, mice were selected for high body weight and low fat content at 42 days for 10 generations. Afterward, mice were monitored for high muscularity by palpation and mice with the highest muscularity on a scale of 1–5 were selected for the next generation. As a result of selection and likely natural mutation(s) over 25 generations, a mouse population with a high muscular phenotype had been generated. A high compact subline was perpetuated through random mating of selected animals (Varga et al. 1997). Sequencing of the myostatin gene in this line revealed a 12-bp deletion (Szabo et al. 1998) leading to a loss of function. Scale and weight-based selection continued for 28 generations. After 86 generations of selection, full sibs with distinct phenotypes were mated. These founder animals became the basis of seven Berlin Muscle Mouse inbred lines (BMMI), four of which carry the *Mstn<sup>Cmpt-dlIAbc* mutation and three are wild type. In this study we used the lines BMMI806 and BMMI816, which are hypermuscular but do not carry the myostatin mutation. At the time of setting up the crossbred experiment, the lines were in generation 21 of inbreeding.</sup>

#### **Pedigree structure**

Two pairs of full sibs of the Berlin Muscle Mouse inbred lines BMMI806 and BMMI816 were crossed reciprocally to generate  $F_1$ ,  $F_2$ , and  $G_3$  intercross populations. For the latter, 93  $F_2$  animals were randomly mated (Schmitt et al. 2009), while avoiding sibling mating, to generate 345  $G_3$  animals.

#### Husbandry and feeding conditions

The mice were treated in accordance with and all experimental protocols were approved by the German Animal Welfare Authorities (approval No. G0405/08). The animals were maintained under conventional conditions at  $22 \pm 2^{\circ}$ C and controlled lighting, with a 12:12-h light:dark cycle. They were kept in groups of two to four animals of the same sex per Macrolon cage and had ad libitum access to food and water. Until the age of 70 days, the animals were fed a standard breeding diet (Altromin standard breeding diet No. 1314 TPF, Lage, Germany). This diet contained 27.0% crude protein, 5.0% crude fat, 4.5% crude fiber, 6.5% crude ash, 50.5% nitrogen-free extract (starch and sugar), vitamins, trace elements, and minerals (2988 kcal/kg metabolizable energy of which 27.0% energy was from proteins, 13.0% from fat, and 60.0% from carbohydrates).

#### Phenotypic measures

After a fasting period of 2 h, 71-day-old mice were anesthetized under isoflurane and decapitated. The Musculus longissimus (ML) and Musculus quadriceps (MQ) were dissected and weighed. The summed muscle weights of the left and right ML and the left and right MQ were recorded as muscle mass (MM). The right muscles were immediately frozen in liquid nitrogen and subsequently stored at  $-80^{\circ}$ C. The left muscles were cooled down to  $6^{\circ}$ C for 1 h, subsequently frozen, and stored at  $-20^{\circ}$ C until WHC and IMF were measured. Carcasses were stored at 6°C and pH values were taken within the *M. biceps femoris* at 1 and 24 h postmortem (ebro PHT 810, Ingolstadt, Germany). For the determination of WHC, frozen muscles were thawed and stored at  $6^{\circ}$ C for 24 h. Muscles were then gently centrifuged for 60 s at  $604 \times g$  in Invitek 1.5-ml receiver tubes with filter inlays to collect the tissue fluid that was not held from the muscle (Eppendorf Minispin, Hamburg, Germany; Invitek, part of STRATEC Molecular GmbH, Berlin, Germany). The ratio of lost tissue fluid to tissue mass was designated as "drip loss." IMF was measured as percentage of muscle weight using nuclear magnetic resonance technology (SMART Trac System, CEM, Kamp-Lintfort, Germany) (Kaerst et al. 2010). Body weights were recorded weekly. At 10 weeks, fat and lean masses were measured by quantitative magnetic resonance (QMR) analysis using the EchoMRI whole-body composition analyzer (Echo Medical Systems, Houston,

TX, USA) (Neuschl et al. 2010; Tinsley et al. 2004). Blood glucose levels were measured after 2 h of fasting, prior to dissection at 10 weeks (Bayer "Contour," Leverkusen, Germany). Muscle glycogen content was determined colorimetrically in the right ML (GOD/PAP method "Glucose liquicolor" by Human, Wiesbaden, Germany), as suggested by Barham and Trinder (1972). Lactate contents were determined colorimetrically in the right ML using the kit by Chung Lee (Lactate Assay Kit, SUNY, Buffalo, NY, USA).

#### Genotyping

Parental BMMI lines were genotyped with the Mouse Diversity Array (Yang et al. 2009) comprising 623,124 single-nucleotide polymorphisms (SNPs). The SNP information provided evidence for allele fixation of 98.3 and 98.0% in lines BMMI806 and BMMI816, respectively, after 21 generations of inbreeding. Both lines differed by 8.8% at the SNP level. Using the information on diverse genomic regions between BMMI806 and BMMI816, 164 informative markers covering all chromosomes (except Y) in an average distance of 16.2 Mb were selected for genotyping the parents, the F<sub>2</sub>, and the G<sub>3</sub> animals (Fig. 1). Regions larger than 10 Mb without informative markers did not differ in SNP alleles between parental lines and were not included in the linkage analysis. Genotyping was done at KBiosciences (Hoddesdon, UK). The genetic map was converted into the physical map using the Mouse Map Converter software from the Jackson Laboratory (Bar Harbor, ME, USA) (Cox et al. 2009). QTL refinement was done by analyzing informative SNPs from the Mouse Diversity Array as described before (Schmitt et al. 2007).

#### QTL and statistical analyses

For QTL analysis, 332  $G_3$  animals were used. Analyses for single-QTL detection and detection of interacting QTL were performed using QTLRel

(http://cran.r-project.org/web/packages/QTLRel/index.html). Unlike the  $F_2$  animals, the  $G_3$  animals may be unequally related to each other. Ignoring the unequal relatedness may result in a serious inflation of false-positive QTL rates. Thus, we applied the mixed model described by Cheng et al. (2010) to analyze our data. All phenotypes were log-transformed to obtain normal distribution. Direction-of-cross and sex were included as additive covariates in the model and used as interactive covariates to test their effects on QTL. Genome-wide significant QTL (P < 0.05) were included in a mixed model for each trait to calculate the respective trait variance in the  $G_3$  population. Trait-specific significance thresholds were estimated from 5,000 permutations (Churchill and Doerge 1994). Additional analyses were performed using R/qtl (Broman et al. 2003). Basic statistics were performed using the SAS software package (SAS System for Windows, Release 9.2, SAS Institute, Cary, NC, USA).

#### Results

#### Phenotypes

Compared to male BMMI806, male BMMI816 showed 46% less total fat mass along with 36 and 22% less IMF in the *M. longissimus* and *M. quadriceps*, respectively. The mean fasting blood glucose levels were also significantly lower in BMMI816 (by 22% in males and 19% in females) than in BMMI806. On average, female BMMI816 mice reached 8% higher body weights and 11% lower muscle masses than BMMI806 females. The analysis of WHC revealed that the MQ of female BMMI816 also showed 51% more drip loss, which indicates a lower WHC than female BMMI806. Sex differences were also distinct between lines. While in line BMMI816 females showed lower WHC and higher fat mass and IMF than males, in line BMMI806 males and females did not differ significantly with respect to WHC and IMF.

When we compared the mean values of the crossbred populations with the parental lines, a dominance effect was observed in BMMI816  $F_1$  males for lower lean mass, fat mass, and IMF, while blood glucose levels, WHC of the ML, and lactate levels were intermediate or resembled the BMMI806 phenotype. In female  $F_1$  animals, even over-dominance effects were found for body weight and lean mass, which were higher, and for IMF of the MQ, which was lower than in the parental lines. The BMMI816 phenotype dominated IMF of the ML, WHC, and blood glucose levels, while the BMMI806 phenotype prevailed for muscle mass, fat mass, and lactate levels in the  $F_1$  females.

In the males of the  $G_3$  population, we found overdominance effects for IMF of the MQ, WHC, muscle mass, and muscle lactate levels with respect to the respective values of the parental lines. The lower BMMI816 phenotypes prevailed for fat mass, IMF of the ML, and blood glucose levels. In female  $G_3$  overdominance was found for body weight and lean mass on the one hand, which were significantly higher, and for IMF and lactate on the other hand, which were lower as compared to both parental lines. The BMMI816 phenotype prevailed by low WHC and blood glucose levels, whereas the BMMI806 phenotype dominated in the higher muscle mass and the lower fat mass in females (Table 1).

In both sexes of the  $G_3$  population, body weight was positively correlated with fat, lean mass, and muscle mass. Significant positive correlations were also found between the IMF of the ML and MQ, but also with total fat mass. Fasting glucose levels were highly dependent on not only body composition but also on the IMF in ML and to a lesser extent in MQ. The higher the IMF was the higher the blood glucose level, which might indicate reduced glucose uptake by the muscle cells and low insulin sensitivity. High fat mass contributed to increased IMF, while high muscle mass (and total lean mass) had positive effects on lowered IMF in both sexes. Interestingly, an opposite correlation between muscle mass and fat mass was observed for both sexes: high muscle mass was negatively correlated with fat mass in males but positively correlated in females. In both sexes, high glycogen content was associated with high drip loss in the ML and MQ, high pH of muscle 1 h postmortem in females, and low pH after 24 h in males (Table 2).

#### QTL effects

Despite the small difference between the parental lines, two QTL for body weight were identified on chromosome (Chr) 1 at 42 Mb (with a LOD support interval between 31 and 51 Mb) and at 71 Mb (59–74 Mb). Animals carrying the BMMI816 alleles showed lower values for this trait. These two QTL for body weight overlapped with QTL for fat and lean mass, since a QTL affecting fat mass was also mapped on Chr 1 at 68 Mb (59–74 Mb) and another QTL for lean mass was mapped on Chr 1 at 38 Mb (23–51 Mb). The alleles for increasing fat mass and lean mass were both inherited from the BMMI806 line. These results are shown in Table 3 and Fig. 2. Furthermore, we identified a QTL for IMF of the ML on Chr 7 at 17 Mb (17–31 Mb), which explains 5.70% of the phenotypic variance in the G<sub>3</sub> population; BMMI806 alleles increased IMF in an additive manner.

Another significant QTL was found for WHC of the MQ on Chr 2 at 134 Mb (128–141 Mb). This QTL accounted for 4.60% of the  $G_3$  variance. The locus showed over-dominance, with heterozygous animals having the highest drip loss (i.e., lowest WHC). For muscle mass, significant QTL were located on Chr 2 at 81 Mb (66–101 Mb) and on Chr 8 at 75 Mb (48–96 Mb). For both QTL, the BMMI816 alleles increased the muscle mass and explained 3.11 and 2.59% of the total phenotypic variance, respectively. QTL on Chr 2 that regulate muscle mass were described before. Their peak positions were located at 76 Mb (64–127 Mb) and 125 Mb (16–161 Mb) (Lionikas et al. 2005). On Chr 8, QTL for muscle mass were detected at 81 Mb (46–86 Mb) (Lionikas et al. 2006) and 110 Mb (27–145 Mb) (Brockmann et al. 2004). An additional QTL for muscle mass was located on Chr 15 at 98 Mb (98–103 Mb). It

was found under the 10% genome-wide significance threshold (LOD = 3.45) with additive and dominance effects of 0.01 and 0.05 g, respectively, and had not described before. No significant QTL for glycogen, lactate, and the variation of pH values were identified.

In addition, we identified two suggestive QTL under a somewhat more liberal 10% genomewide significance threshold (P < 0.1). For fasting blood glucose levels, we found a suggestive QTL on Chr 1 at 68 Mb (55–77 Mb, LOD = 3.40). Consistent with the parental phenotypes, the additive effect of the BMMI806 allele for blood glucose levels was 4.7 mg/ dl (dominance effect = 0.8 mg/dl). This QTL coincided with the QTL interval for fat mass and body weight.

Moreover, a significant QTL for the pigmentation of the BMMI816 was identified on Chr 7 at 97 Mb (94–98 Mb). The BMMI816 line showed black pigmentation except for the ventral side of the body, while the BMMI806 animals were not pigmented. Those  $G_3$  animals that were homozygous or heterozygous for the BMMI816 allele at this locus exhibited the specified pigmentation.

#### Informative loci

Additional QTL refinement was accomplished when we analyzed the degree of divergence between the parental BMMI lines. The extent of divergence and the number of genes in these informative loci are given in Table 3, according to the data of all SNP markers from the Mouse Diversity Array. A list of putative candidate genes, which are located in the divergent regions, is given in Table 4A and B for the respective examined traits. However, albeit a high SNP density on the array, it is very likely that not all divergent loci were detected. Therefore, it cannot be excluded that genetic determinants underlying certain QTL effects are located between the visible divergent regions. Additional data, e.g., sequencing and RNA expression, are needed to verify candidate genes that were obtained from this analysis. Using rs-SNP marker information of the Mouse Diversity Array and a perl script, we generated Fig. 3, which illustrates the ratio between divergent regions and regions identical by descent (IBD) between the two parental lines for the detected QTL (Mott 2005). The regions that are divergent between the two parental strains most likely contain the genes responsible for the detected QTL effects. The close relatedness of the parental lines explains the high amount of IBD haplotypes.

#### Discussion

The present study was performed to identify QTL for muscle traits, with IMF and WHC being the main traits of interest. It was unclear whether IMF-influencing genes would be different from genes responsible for the distribution and quantity of normal adipose tissue (Tanomura et al. 2002). From our results that showed different QTL for IMF and total fat mass, we conclude that there are indeed different genetic determinants that control total fat mass and IMF. The highly significant QTL for total fat mass on Chr 1 in our cross was also seen in the scans for IMF of the *M. longissimus* but had no significant effects (P < 0.37) and did not affect IMF in the *M. quadriceps*. This QTL likely acts predominantly on fat accumulation in the adipose but not in the muscle tissue. The total fat mass QTL on Chr 1 (Fig. 1) overlapped with a suggestive QTL (P < 0.10) for fasting blood glucose levels. It cannot be excluded that this chromosomal region is a major regulator of both glucose utilization and fat metabolism in our cross. It has been shown repeatedly that increased adiposity is linked to insulin resistance and higher blood glucose levels as a result of impaired glucose clearance (Kahn et al. 2006). In addition, high intramuscular triglyceride levels, which are highly correlated with high body fat mass in our cross, inhibit the insulinsignaling cascade and lower glucose uptake (Chadt et al. 2008; Powell et al. 2004; Schmitz-Peiffer 2000; Tanomura et al. 2002). Both high fasting blood glucose levels and increased

IMF were observed in BMMI806 mice. Different genes in the QTL region could contribute to the QTL effect. One interesting candidate gene in the QTL region on Chr 1 at 67 Mb is the long-chain acyl-CoA-dehydrogenase gene (*Acadl*), which is directly involved in the first step of  $\beta$  oxidation by converting acyl-CoA from fatty acids into  $\Delta^2$ -trans-enoyl-CoA. Decreased oxidation of mitochondrial fatty acid was shown in *Acadl*-deficient mice (Zhang et al. 2007). It seems possible that impaired fatty acid utilization could result in increased fat mass but also in high IMF, especially since skeleton muscles lack de novo lipogenesis (Eaton 2002). Interestingly, data from the Mouse Diversity Array showed three SNPs to be different in this gene between BMMI806 and BMMI816 mice. The first SNP is located in an intron and the second and the third SNPs are located within exons and code synonymously. However, since it cannot be ruled out that adjacent base pairs are also different leading to a nonsynonymous coding, *Acadl* seems to be a good candidate gene that deserves further attention.

As mentioned, it is known from studies in humans and mice that muscular triglycerides and certain lipid species inside the muscle fibers have similar metabolic effects, e.g., on the glucose uptake (Ebeling et al. 1998; Pan et al. 1997; Phillips et al. 1996; Yu et al. 2002). To some extent, the biological function of muscle lipid stores seems to be similar in different species in terms of lipids as an energy fuel and their role in homeostasis. However, the respective patterns of fatty acids might be slightly different in different species and can even be manipulated (Ludden et al. 2009; Wood et al. 2004). In our study we measured the total fat content within the muscle and found differences between parental mouse lines.

Several fat metabolism genes are located in the QTL region on Chr 7 that affects IMF, e.g., the lipase gene (*Lipe*) and the apolipo-protein E gene (*Apoe*) (Hansson et al. 2005; Hunt et al. 2006), which might be responsible for the high IMF phenotype of the BMMI806. However, both genes are located in regions that are noninformative on the measured SNP level.

Still, there might be sequence alterations in these regions that remained undetected by the Mouse Diversity Array. Another interesting candidate gene in this region is thymoma viral proto-oncogene 2 (*Akt2* or *Pkb*) at 28 Mb. *Akt2* is involved in the insulin-dependent regulation of lipid metabolism and triglyceride storage (Leavens et al. 2009). There is a SNP marker in this gene that differs between the parental lines, and the BMMI806 animals have high fasting blood glucose levels that might also indicate a certain insulin resistance. This shows that including information about regions that are identical by descent between the parental lines can help to narrow down the list of candidate genes in a QTL, but additional data, such as from RNA expression or sequencing, are needed to finally confirm candidate genes (Table 4A, B).

The comparison of syntenic regions between mice and other mammals via the Ensembl database revealed supporting evidence for the presented data. The murine QTL region for IMF on Chr 7 is syntenic with regions on the *Sus scrofa* chromosomes (SSC) SSC6 (25–43 Mb) and SSC14 (138–140 Mb) (Birney et al. 2004; Rohrer et al. 1996). In the syntenic regions on SSC6, one QTL for IMF and five additional fat-related QTL have been identified in pigs (de Koning et al. 2000; Liu et al. 2008; Paszek et al. 2001), while the small region on SSC14 is located in a region where two fat-related QTL have been found (Dragos-Wendrich et al. 2003; Knott et al. 1998). Therefore, it seems possible that genes in these syntenic regions have similar functions in both species. The presented mouse model, which is simplified in its genomic structure, could support the identification of genes that might affect IMF in pigs.

The second objective of this study was the identification of QTL for WHC. One QTL for this trait that was specific for the *M. quadriceps* was identified on Chr 2. The *M. longissimus* did not show differences in WHC between the parental lines. The particular fiber type composition of the two different muscles could be responsible for differences in WHC and IMF. Depending on the ratio of oxidative and glycogenic fiber types, the metabolism of the tissue can be different (Armstrong and Phelps 1984; Hamalainen and Pette 1993; Rehfeldt et al. 2010). Moreover, a correlation existed between glycogen and WHC, since higher glycogen content was also associated with higher drip loss. It is known that high glycogen storage impairs the WHC of the tissue. With postmortem hypoxia, glycogen degradation results in an increased accumulation of lactate, which among other factors lowers cellular pH values. The timing and extent of this lactate accumulation depend on the amount of stored glycogen. The concomitant pH drop might alter the structural integrity of the muscle fibers' cellular scaffold and membrane proteins, which results in an increased loss of cell form and cell fluid (Bee et al. 2007; Choe et al. 2008). The parental BMMI816 line showed higher glycogen content and inferior WHC.

For the lactate content of the *M. longissimus*, a significant, positive correlation was found with the WHC in the *M. longissimus* in males, and with the pH 24 h after dissection in females. Since the *M. longissimus* samples, in which the glycogen and lactate contents were measured, were taken immediately after exsanguinations, it is possible that hypoxia did not last long enough to observe stronger correlations of lactate and glycogen with carcass pH. However, the results yet support the model of postmortem scaffold destruction and concomitant loss of tissue fluid due to increased tissue acidity that is influenced by the levels of muscular glycogen and lactate at the time of death. For the QTL for WHC on Chr 2 there was a syntenic region on SSC17 (13–37 Mb) in which no WHC-related QTL was found.

Comparative genomics and interspecies research are used to transfer the information obtained from our mouse study into the field of livestock research to which it is supposed to contribute. Using syntenic regions in the context of QTL studies can thereby provide additional information about the plausibility of identified QTL. However, it must be considered that the chance of finding syntenic regions between mouse and pig on a chromosome that contains any QTL is high, depending on the number of QTL and the size of the respective QTL intervals in both species. This is the case, e.g., for IMF and WHC (as drip loss), for which 15 of the 19 porcine chromosomes contain several significant QTL. Therefore, chances are relatively high to find a OTL for IMF on any of the pig chromosomes. However, for the murine IMF QTL on Chr 7 (17-31 Mb), only two syntenic regions exist, one on SSC6 (25-43 Mb) and one on SSC14 (138-140 Mb), but only the region on SSC6 contains a QTL for IMF. More than 50% of all murine genes are identified as having orthologs in pigs, which does not prove but indicate many putative similar gene functions (BioMart, Ensembl). Comparing mouse data to the results from pigs contributes additional information not only about orthology but also about the actual functions of conserved genes in both species.

Both mouse lines used in this study are hypermuscular and share large portions of their genomes. Nevertheless, they show large differences in their phenotypes. The increased body weights of the BMMI lines reflect the selection response during the historic breeding process. It appeared reasonable to examine these lines as models for livestock, since commercially used races of pigs or cattle are generally larger than their wild ancestors, as are the BMMI lines compared to wild-type mice. By generating a G<sub>3</sub> population, we expected narrower QTL intervals but had to use a relatively more complicated analysis. Breeding additional generations will probably enable us to fine map these regions better. The QTL identified in the G<sub>3</sub> in this study provide evidence to look for interesting alterations in the BMMI genome that explain the different phenotypes. Since the lines are

related and have most haplotypes in common, we also expect fewer differences in the DNA sequence between the lines as compared to unrelated lines. This circumstance also could decrease the number of putative candidate genes in subsequent sequencing and expression studies.

#### Conclusion

The Animal Genome database (animalgenome.org) reports 54 and 37 significant QTL for IMF and WHC, respectively, but only a few candidate genes have been identified in pigs so far. Use of our inbred mouse lines in a controlled environment with the whole range of tools that are available for mice (with their long history of being genetic models) and their short generation interval might contribute to increasing the knowledge about livestock genetics for the examined traits. However, the success of cloning QTL in mice was quite limited in the past. The use of outbred populations seems to be more effective for identifying causative mutations (Brockmann and Bevova 2002; Valdar et al. 2003). Using advanced intercross lines (AIL) seems to be beneficial in this regard compared to the classic  $F_2$  approach (Darvasi and Soller 1995). The AIL approach leads to more recombination events and smaller confidence intervals, while the close relatedness between the inbred parental lines initially decreases the number of regions that probably harbor phenotype-affecting alterations in the genome. We took this into account by using a G<sub>3</sub> instead of an F<sub>2</sub> population. Analyses of RNA expression and sequencing based on the identified QTL region should allow the identification of the causative mutation(s) for the examined traits in the near future.

To our knowledge this is the first report on QTL mapping for IMF, WHC, and associated traits in mice. The localization of different genomic regions affecting IMF and WHC yielded information for fine mapping these regions and identifying putative candidate genes. The identified loci partly coincide with QTL for these traits in other species, especially pig. Therefore, the BMMI model is an interesting genetic resource not only to identify genes, which could also contribute to QTL effects in other species, but also to test identified candidate genes for their functions in genetically modified mice.

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#### Fig. 1.

MapChart plot of 164 reference single-nucleotide polymorphisms used in this study. Positions are given in Mb (Ensembl release 37). *Bars* indicate 1-LOD support intervals of identified significant QTL



#### Fig. 2.

Effect plots showing means and standard errors of the three genotype classes at all significant QTL peak positions for males (*filled circles*) and females (*open circles*); QTL significant under a P < 0.05 and **b** P < 0.10 genome-wide threshold; 816 homozygous BMMI816 allele, 806 homozygous BMMI806 allele, H heterozygous animals

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informative rs-SNP marker	BMMI806 BMMI816 ide	entical by descent
Lean mass Chr1 (23 - 51 Mb)		
Fat mass Chr1 (59-74 Mb)	WHC-MQ Chr2 (128-141 Mb)	IMF-ML Chr7 (17-31Mb)
Muscle mass Chr2 (66-101 Mb)		
Muscle mass Chr8 (48-96 Mb)		

#### Fig. 3.

Illustration of divergent genomic regions and regions being identical by descent between BMMI806 and BMMI816 mice for different quantitative traits, based on significant QTL and respective rs-SNP marker information

Table 1

Body traits of parental, F<sub>1</sub>, and G<sub>3</sub> animals

Trait	Р				$\mathbf{F}_{\mathbf{I}}$		G3	
	BMMI816		BMMI806		Males	Females	Males	Females
	Males	Females	Males	Females				
Body weight (g)	36.88 (3.68)	29.02 (2.77)**	36.43 (4.69)	26.77 (2.16)	35.70 (4.21)	30.73 (2.69) <sup>a,bbb</sup>	38.99 (3.18)	31.60 (2.72) <sup>aa,bbb</sup>
Lean mass (g)	31.09 (2.19)	23.04 (2.02)	29.31 (2.29)	23.87 (1.99)	30.84 (3.55) <sup>bb</sup>	26.00 (2.74) <i>aa</i> , <i>b</i>	32.56 (2.58) <sup>bbb</sup>	25.94 (2.16) <sup>aa,bb</sup>
Muscle mass (g)	1.14(0.10)	$0.78~(0.09)^{*}$	1.12 (0.12)	0.88 (0.07)	1.14 (0.22)	0.93 (0.12) <i>aaa</i>	$1.25\ (0.14)^{aa,bbb}$	0.96 (0.10) <i>aaa</i>
Fat mass (g)	2.37 (0.65) <sup>***</sup>	4.26 (1.42)	4.41 (1.46)	3.13 (1.19)	$2.00(0.91)^{bbb}$	2.28 (0.63) <sup>a</sup>	2.50(1.32)bbb	2.42 (1.03) <i>aa</i>
IMF-ML (%)	$1.68 (0.30)^{***}$	2.07 (0.42)	2.64 (0.34)	2.45 (0.49)	$1.69~(0.37)^{bbb}$	1.84 (0.57) <sup>bbb</sup>	$1.62 (0.39)^{bbb}$	1.58 (0.39) <sup>aaa,bbb</sup>
IMF-MQ (%)	$1.41 (0.20)^{***}$	1.94(0.17)	1.81 (0.30)	1.80 (0.29)	1.41 (0.23) bbb	1.58 (0.32) <sup>aaa,bb</sup>	1.28 (0.19) <sup>a,bbb</sup>	1.32 (0.19) <sup>aaa,bbb</sup>
WHC-ML (%)	0.92 (0.25)	1.29 (0.35)	1.09 (0.37)	1.01 (0.38)	$1.29 (0.64)^{a}$	$1.36\ (0.51)^b$	1.38 (0.47) <i>aaa,bb</i>	$1.48~(0.43)^{bbb}$
WHC-MQ (%)	0.89 (0.27)	$1.22~(0.34)^{*}$	0.85 (0.33)	0.81 (0.26)	1.15 (0.65)	$1.18(0.51)^{bb}$	$1.19~(0.42)^{aa,bbb}$	$1.16(0.35)^{bbb}$
Glucose (mg/dl)	91 (17) <sup>***</sup>	84 (13) <sup>***</sup>	120 (26)	104 (17)	$106(26)^{a,bb}$	86 (18) <i>bbb</i>	96 (17) <i>bbb</i>	83 (12) <i>bbb</i>
Glycogen (mg/g)	2.78 (0.89)	2.83 (1.68)	2.47 (0.76)	2.45 (1.38)	$1.56(0.87)^{d}$	1.74 (1.62)	1.82 (0.80)	1.81 (0.73)
Lactate (mg/g)	0.08 (0.01)	0.08 (0.02)	0.06 (0)	0.07 (0.03)	$0.06\ (0.01)^{d}$	$0.05 (0.01)^{a}$	0.05 (0.03) <i>aaa,bbb</i>	$0.04 \ (0.02)^{a,bbb}$

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IMF intramuscular fat content, WHC water-holding capacity (measured as drip loss in percent of muscle weight), ML Musculus longissimus, MQ Musculus quadriceps, Glucose blood glucose level

Phenotypes of the parental lines are compared with each other and the F1 and G3, separated for sex. Asterisks indicate significant differences between the parental lines; letters indicate significant differences between the  $\mathrm{F1}$  or G3 and BMMI816

<sup>a</sup> or BMMI806

 $\boldsymbol{b}$  lines. Number of asterisks and letters show levels of significance

\*\*\*/aaa/bbbP < 0.001,

 $^{**/aa/bb}_{P < 0.01}$ ,

 $^{*/a/b}P < 0.05)$ 

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Table 2

Spearman's correlation coefficients between different traits in the G<sub>3</sub> population

							Males						
	Body weight (g)	Lean mass (g)	Fat mass (g)	Muscle mass (g)	IMF-ML (%)	IMF-MQ (%)	WHC-ML (%)	WHC-MQ (%)	Glucose (mg/dl)	pH1	pH24	Glycogen (mg/g)	Lactate (mg/g)
Body weight (g)		$0.91^{***}$	$0.34^{***}$	$0.64^{***}$	0.09	0.06	0.02	-0.05	$0.24^{***}$	-0.01	0.03	-0.10	0.13
Lean mass (g)	$0.90^{***}$		0.01	$0.78^{***}$	-0.15	-0.18	0.05	0.00	$0.22^{**}$	-0.08	0.04	-0.03	0.11
Fat mass (g)	$0.42^{***}$	0.11		-0.18	$0.71^{***}$	$0.70^{***}$	-0.07	-0.08	$0.17^{*}$	$0.20^{**}$	-0.07	-0.15 *	0.05
Muscle mass (g)	$0.68^{***}$	$0.71^{***}$	$0.17^{*}$		-0.24 **	-0.26	0.03	-0.05	0.13	-0.05	0.07	-0.14	0.11
IMF-ML (%)	0.02	-0.12	$0.50^{***}$	-0.15		$0.67^{***}$	-0.03	-0.09	0.10	$0.16^*$	$-0.15^{*}$	-0.15	-0.04
IMF-MQ (%)	-0.10	-0.25 ***	0.45***	-0.23 **	$0.49^{***}$		-0.12	-0.03	0.05	0.02	-0.26 ***	-0.13	0.03
WHC-ML (%)	-0.03	-0.03	0.07	0.04	0.11	0.01		$0.41^{***}$	-0.02	0.10	-0.15*	$0.36^{***}$	$0.16^*$
WHC-MQ (%)	-0.02	-0.03	0.00	-0.07	0.03	0.03	0.14		-0.06	-0.02	-0.24 ***	$0.37^{***}$	0.04
Glucose (mg/dl)	0.15	$0.17^{*}$	0.10	0.04	$0.17^{*}$	0.08	0.13	0.10		0.10	$0.17^{*}$	0.12	0.00
pH1	$0.17^{*}$	0.11	0.10	$0.23^{**}$	-0.06	-0.06	-0.01	0.02	0.08		$0.34^{***}$	0.02	0.07
pH24	0.16	0.08	0.07	0.14	-0.06	-0.06	-0.06	-0.21	0.01	$0.25^{**}$		-0.26 ***	0.01
Glycogen (mg/g)	-0.01	0.01	-0.02	0.14	-0.05	-0.04	$0.17^{*}$	$0.22^{**}$	0.03	$0.21^{**}$	-0.18*		0.01
Lactate (mg/g)	-0.01	-0.04	0.00	-0.03	-0.04	-0.14	0.08	-0.04	0.01	-0.01	$0.16^*$	-0.02	
Females													
<i>IMF</i> intramuscular fa	at content, WHC wat	ter-holding capacit	ty, measured as dr	rip loss in percent of	muscle weight, <i>M</i>	IL M. longissimus,	MQ M. quadriceps	i, <i>Glucose</i> blood glu	icose level				
Asterisks indicate the	significance of corr	relations:											
*** P < 0.001,													
$^{**}_{P < 0.01}$													
$^{*}_{P < 0.05}$ ,													
$\bullet$ $P < 0.10$													

# Table 3

QTL significant at the genome-wide level of 5% for various traits in the G<sub>3</sub> population at the age of 10 weeks

Trait	Chr	Mba	qIS	Marker <sup>c</sup>	r0Dd	a (SE) <sup>e</sup>	d (SE) <sup>e</sup>	$\%~{ m G}_3~{ m var}^f$	iSNP%	#genesh
Lean mass (g)	-	38	23-51	rs4222269	3.66	-0.83 (0.22)	-0.36 (0.29)	2.22	20.1	71
Body weight (g)	-	42	31–51	rs4222320	6.46	-1.41 (0.31)	-0.27 (0.36)	4.21	19.1	65
Fat mass (g)	1	68	59–74	rs31886089	7.02	-0.53 (0.11)	-0.27 (0.13)	9.84	14.6	35
Body weight (g)	-	71	59–74	rs31886089	7.99	-1.54 (0.34)	-0.69 (0.39)	5.32	14.6	35
Muscle mass (g)	2	81	66-101	rs13476603	4.29	0.06 (0.01)	0.03 (0.02)	3.11	16.1	92
WHC-MQ (%)	2	134	128–141	rs27268102	3.93	0.06 (0.03)	0.17 (0.04)	4.60	20.4	30
IMF-ML (%)	٢	17	17–31	rs3675839	3.72	-0.13 (0.03)	-0.06 (0.04)	5.70	4.4	37
Muscle mass (g)	×	75	48–96	rs31860676	3.63	0.05 (0.01)	0.02 (0.01)	2.59	6.3	51

<sup>u</sup>Most likely chromosomal location given as Mb position

 $b_{1-LOD}$  support interval (SI) in Mb

 $^{\ensuremath{\mathcal{C}}}$  Marker closest to the chromosomal position with the highest LOD score

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 $^{d}$ LOD scores from the full model estimations

 $^{e}$ Additive (a) and dominance (d) effect and their standard errors (SE) determined with the nontransformed raw trait values, therefore given in the respective unit; the direction of a and d, respectively given as the effect of the BMMI816 allele

 $f_{
m G3}$  phenotypic variance (%) explained by the QTL; QTL effect given as the reduction of the residual sum of squares fitting 1 vs. 0 QTL

 $^{\mathcal{S}}$  Percentage of informative SNP markers in the LOD support interval

 $\hbar_{\rm N}$  Number of genes that are located in informative loci within the SI

#### Table 4

List of candidate genes for different traits in informative regions of the QTL

<u>A</u>				
Lean mass (Chri	1:23–51 Mb)			
1110058L19Rik	Cnga3	Inpp4a	Tesp1	
1700024G10Rik	Cnnm3	Khdrbs2	Tesp2	
2010300C02Rik	Cnnm4	Lincr	Tmem131	
2300002018Rik	Col19a1	Lman2l	Tpp2	
4832428D23Rik	Col5a2	Lmbrd1	Tsga10	
4921511C04Rik	Creg2	Lonrf2	Txndc9	
4921533L14Rik	D1Bwg0212e	Map4k4	Ugcgl1	
6330578E17Rik	Dnahc7	Mgat4a	Unc50	
9430069J07Rik	Dnahc7b	Mitd1	Uxs1	
A230074B11Rik	Dst	Mrpl30	Wdr75	
Actr1b	EG212225	Nms	Zap70	
Aff3	Eif5b	Npas2		
Arid5a	ENSMUSG00000053185	Pdcl3		
Bai3	Ercc5	Phf3		
BC043098	Fhl2	Plekhb2		
BC050210	Gm776	Prim2		
Bivm	Gulp1	Rev1		
C230029F24Rik	1118r1	Slc39a10		
Cfc1	<i>ll1r2</i>	Slc40a1		
Chst10	Il1rl2	Tbc1d8		
Fat mass (Chr1:59	9–74 Mb)	IMF-ML (Chr7:1	7–31 Mb)	
1110028C15Rik	Klf7	1700049G17Rik	Pvrl2	
9430067K14Rik	March4	4933426121Rik	Ryr1	
A830006F12Rik	Mpp4	Actn4	Samd4b	
Abi2	Mreg	Akt2	Sertad3	
Acadl	Mtap2	BC057627	Sfrs16	
Als2cr13	Nbeal1	Bckdha	Shkbp1	
BC042720	Pard3b	Blvrb	Sipa113	
Bmpr2	Pecr	Ceacam13	Spnb4	
C030018G13Rik	Pip5k3	Ckm	Sympk	
Carf	Pthr2	Cyp2b10	Tex101	
Cd28	Spag16	Cyp2s1	Trappc6a	
Cps1	Tmem169	Ercc2	Xrcc1	
Cyp20a1	Wdr12	Gemin7	Zfp30	
D630023F18Rik	Xrcc5	Grik5	Zfp568	
Erbb4	Zdbf2	Nlrp4e	Zfp60	
Gm973		Nlrp5	Zfp607	

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A		
Lean mass (Chr1:23–51 Mb)		
Icall	Numbl	Zfp84
Icos	Pou2f2	
Idh1	Prx	
Ikzf2	Ptgir	
WHC-MQ (Chr2:128-141 Mb)		
1110034G24Rik	Plcb1	
2210009G21Rik	Prei4	
2310003L22Rik	Sel112	
2900006F19Rik	Smox	
4921508D12Rik		Snap25
4930473A02Rik		Snrpb
Anapc1		Sptlc3
Atrn		Tasp1
Btbd3		Txndc13
Cdc25b		Zc3h8
Cds2		
Ebf4		
Esfl		
F830045P16Rik		
Flrt3		
Hao1		
Jagl		
Mertk		
Mkks		
Pdyn		

Muscle mass (Ch	r2:66–101 Mb)
1110051M20Rik	Cugbp1

В

1110051M20Rik	Cugbp1	Madd	Olfr1258	Syt13
2700094K13Rik	D030051N19Rik	Metapl1	Olfr1271	Tfpi
4833418A01Rik	D230010M03Rik	Mtch2	Osbpl6	Tnks1bp1
4833423E24Rik	D2Ertd391e	Myo3b	P2rx3	Ttc21b
4932414N04Rik	D430039N05Rik	Nckap1	Pde1a	Ube2e3
Abcb11	Ddb2	Ndufs3	Pex16	Ube2l6
Acp2	Dnajc10	Nup160	Phf21a	Xirp2
Agbl2	Dusp19	Nup35	Ppp1r1c	Ypel4
Agtrl1	ENSMUSG0000075313	Olfr1032	Prdx6_rs1	Zc3h15
Arhgap1	Ext2	Olfr1033	Prg2	Zdhhc5
B3galt1	F2	Olfr1107	Psmc3	Zfp533
BC003993	Hsd17b12	Olfr1115	Ptprj	Zfp804a
Calcrl	Itga4	Olfr1182	Rapgef4	

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B			
Muscle mass (Ch	r2:66–101 Mb)		
Chn1	Itgav	Olfr1193	Scn1a
Chst1	Kbtbd4	Olfr1206	Scn7a
Ckap5	Lass6	Olfr1213	Slc43a1
Clp1	Lnp	Olfr1215	Smtnl1
Creb3l1	Lrp2	Olfr1219	Ssfa2
Cry2	Lrp4	Olfr1221	Ssrp1
Ctnnd1	Lrrc55	Olfr1245	Stk39
Muscle Mass (Chr	8:48–96 Mb)		Muscle Mass (Chr15:98-103 Mb
1700007B14Rik	Es22	Sh2d4a	Zfp641
1810029B16Rik	Frem3	Sh3rf1	Pou6f1
2810422J05Rik	Fto	Slc10a7	Adcy6
4732435N03Rik	Gab1	Stox2	Slc4a8
Abcc12	Gatad2a	Tktl2	Cacnb3
Adcy7	Inpp4b	Tmem188	Scn8a
Aga	Large	Tox3	Ccdc65
AI931714	March1	Ttc29	Espl1
Arhgap10	Nanos3	Vegfc	Atp5g2
Atp6v1b2	Nat2	Wdr17	Mll2
BC033932	Nat3	Zfp423	Spats2
BC053440	Nek1		Lima1
BC057552	Nkd1		
Brd7	Nod2		
Cyp4f18	Nr3c2		
D630040G17Rik	Odz3		
Ednra	Otud4		
Ell	Palld		
EG330776	Psd3		
EG70793	Rab8a		
Pigmentation (Chi	7:94–98 Mb)		
Nox4			
Tyr			
Grm5			
Ctsc			
Rab38			
Me3			
Svtl2			