

1 **A comparison of methods to pre-adjust data for systematic environmental**
2 **effects in genetic evaluation of sheep**

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9 Short title: Adjustment of data prior to genetic evaluation of sheep

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25

1 **Abstract**

2
3 Sheep performance data are routinely adjusted to account for systematic environmental effects
4 but little attention has been given to the choice between the methods available. This study
5 compares methods to adjust liveweight measurements for individual age (within group
6 regression versus age intercept) and age of dam (dam age class versus quadratic polynomial on
7 dam age) and fat and eye muscle depth for liveweight (within group regression versus quadratic
8 polynomial on weight). Criteria for comparison included: 1) changes in phenotypic variance
9 and heritability estimates; 2) predictive ability of resulting estimated breeding values (EBVs);
10 3) correlation between adjusted and adjustor variables; and 4) ease of computation. With the
11 exception of dam age adjustment of liveweight, adjustment methods did not influence the
12 phenotypic variance of the trait being adjusted. Increases in the direct heritability were observed
13 for liveweight adjusted using the age intercept, dam age class and quadratic methods, and fat
14 depth using quadratic adjustment for liveweight. Within group regression based adjustments
15 generally resulted in a reduction of direct heritability. Maternal heritability and permanent
16 environment due to dam were unaffected by the adjustment methods utilised. The regression of
17 progeny performance on sire estimated breeding value also indicated that the age intercept
18 produced estimated breeding values that were closer to the expectation than those using the
19 within group regression adjustment. The age intercept and quadratic polynomial methods of
20 adjustment appear to be slightly more appropriate for genetic evaluation purposes.

21 **Keywords:** Data adjustment, sheep, liveweight, age, dam age, fat depth, eye muscle depth

23 **Introduction**

24
25 Accurate genetic evaluation of animals requires data to be corrected to accommodate differences
26 in known systematic environmental effects that influence the growth and performance of
27 livestock. These effects usually relate to differences between animals within contemporary
28 groups (CG) that influence phenotypic performance but are not a direct result of genetic
29 differences between these animals. Adjustment of performance records for known environmental
30 effects aims at reducing the non-genetic or environmental components of phenotypic variance
31 (Raymond 1982). In the lamb industry examples include age of the animal, age of the animal's
32 dam, birth type and rearing type.

33
34 The methods used to adjust data for these effects are varied and have been compared by
35 Raymond (1982). Methods of adjustment can be classed as additive, where a pre-determined

1 number of units of the trait are added for each unit of time or weight, or multiplicative in which
2 the observation is increased or decreased on a percentage basis. Recent genetic evaluation has
3 utilised five main methods of adjustment: 1) Simple multiplicative factors calculated from class
4 means; 2) linear regression; 3) ordered polynomials which work additively; 4) additive class
5 effects; and 5) age intercept adjustment which is a multiplicative adjustment.

6
7 A regression of liveweight (LWT) on age has been commonly used to adjust LWT to a
8 standard age. This is achieved by using the within group relationship between age and weight.
9 Within group regression adjustments are additive type adjustment methods. Instead, the age
10 intercept method is a multiplicative adjustment method based on extrapolation from a straight
11 line that runs from the age-weight record to an age-intercept on the x-axis corresponding to
12 zero weight. The age intercept methods are routinely used in BREEDPLAN (Johnston *et al.*
13 1999) and now in OVIS (Brown *et al.* 2000) for national genetic evaluation of beef cattle and
14 sheep, respectively. The reasoning behind the age intercept adjustment is that it allows
15 differences in weights across animals within a CG to get larger or smaller over time, as would
16 be biologically anticipated. The age-intercept is calculated by estimating the linear
17 relationship between liveweight and age over all animals after accounting for fixed and
18 random effects. The equation for this relationship is then back-solved to obtain the value
19 where weight is equal to zero (x axis intercept).

20
21 A number of other methods for correcting liveweight for age have been investigated. These
22 include pooled regression (within sire groups), average daily gain and least squares (Gregory
23 *et al.* 1976; Gregory *et al.* 1977a). The method used to correct liveweight observation for age
24 must also be capable of eliminating variations in weights caused by non-random use of sires
25 during the mating period (Gregory *et al.* 1977b). Non-random use of sires can lead to age
26 being heritable (Gregory *et al.* 1977b). For example if a particular sire is consistently used in
27 as a back-up sire in an AI program all of the progeny of this sire are going to be above
28 average. Given the likelihood of these mating structures in today's sheep flocks this is an
29 important consideration.

30
31 The influence of age of dam on progeny performance has been identified in most livestock
32 species (Gregory *et al.* 1977c; Wilson *et al.* 1996; Bourdon, 1997, pp. 170-173). Data are
33 commonly adjusted to a standard age when dams are at peak production. If the dam is
34 younger or older than this standard age the LWT will be adjusted up to compensate
35 accordingly. Animals can be allocated to dam age classes and their data adjusted using a

1 simple additive adjustment factor for the respective dam age class. Alternatively, one could
2 use higher order polynomial adjustments. BREEDPLAN and OVIS routinely use quadratic
3 functions to multiplicatively adjust LWT for dam age effects.

4
5 Strong phenotypic and genetic relationships exist between age, LWT, fat depth (FATD) and
6 eye muscle depth (EMD) (Atkins *et al.* 1991; Brash *et al.* 1992; Gilmour *et al.* 1994). As a
7 result, FATD and EMD are commonly adjusted for differences in age or LWT at
8 measurement to compensate for these relationships. In beef cattle, heritability estimates for
9 carcass traits tend to be higher when adjusted for weight rather than age (Meyer 1999). The
10 two main methods used are regression of FATD and EMD on LWT within CG and quadratic
11 adjustment. These two methods are similar to those described above.

12
13 Optimal selection of adjustment methods should involve the comparison of actual and
14 adjusted observations with the adjuster variables, variances of the actual observations and
15 adjusted observations and the effects of the adjustment factors on the heritability estimates
16 (Raymond 1982). Further, the predictive ability of EBVs as computed from the regression of
17 progeny performance of parents EBVs should also help identify the optimum adjustment
18 methods.

19
20 Based on these criteria, the aim of this paper is to compare a number of methods used to adjust
21 performance data for systematic environmental effects due to individual age, dam age and LWT
22 prior to genetic evaluation.

23 24 **Materials and Methods**

25
26 Post-weaning data on LWT and ultrasound scans of FATD and EMD were retrieved from the
27 White Suffolk breed in the LAMBPLAN (Banks 1994) database, based on pedigree known and
28 observations recorded. Initial edits removed records beyond 3 standard deviations within CG for
29 age, LWT, FATD and EMD. Also, only those animals with dams younger than 10 years of age
30 and from CGs with more than 40 animals were included. This editing resulted in 8418 animals
31 being in the data file with 164 CGs and 19236 animals in the pedigree file. There were 5411
32 dams and 539 sires with progeny with data. These data were used to calculate adjusted
33 observations using two methods each for individual age and age of dam adjustment of LWT,
34 and FATD and EMD adjustment for LWT.

Adjustment of LWT for individual age

Regression

LWT observations were adjusted using a linear regression of LWT on age for all animals within a CG. For the *i*th CG this regression coefficient was calculated from a weighted average using the following formula:

$$\frac{(N_s \times b_s) + (N_i \times b_i)}{N_s + N_i},$$

where N_s = number of animals in a standard CG,

N_i = number of animals in the *i*th CG,

b_s = standard regression of weight on age,

b_i = actual regression of weight on age for the *i*th CG.

Hence the influence of the standard regression coefficient will diminish as the size of the *i*th CG increases. The standard CG size was the most common (mode) CGs size in the data file. This regression coefficient was then used to adjust all LWT measurements to the average age of all animals in that CG.

Age Intercept

Adjusted LWT using the age intercept method was calculated using the following formula:

$$Y_a = Y_o (A_a - A_i) / (A_o - A_i)$$

Where: Y_a = Adjusted weight,

Y_o = Observed weight,

A_a = Standard age,

A_i = Age intercept corresponding to zero weight,

A_o = Observed age.

Expectation of adjusting LWT for age using these methods

To illustrate how within group regression and age intercept adjustment work a data set was deterministically simulated using SAS (1990). This data set consisted of 180 animals ranging in age from 100 to 280 days. Animals were allocated to three CGs based on age. Liveweight were assigned using the following growth function, $wt=70*(1-0.93e^{-0.005Age})$, which resulted in similar weights at the start, middle and end of the growth period to those in the data used in this

1 study. Adjustment factors were then estimated using SAS (SAS 1990) and applied to these
2 fabricated data. Results are illustrated in Figure 1.

3
4 [Please insert Figure 1 near here]

5
6 Both the within group regression and age intercept adjustment methods nullify the relationship
7 between age and liveweight within groups. However over all data there is a positive relationship
8 between within group regression adjusted liveweight and age.

9
10 ***Adjustment of LWT for age of dam***

11
12 *Dam age classes*

13
14 Animals were classified into three dam age classes: 1 to 2 years, 3 to 5 years and 6 to 10 years
15 old. Animals within each of these classes received a simple multiplicative adjustment. For
16 example, if the estimated adjustment factor for dam age class 1 is 1.02 then all animals with a
17 dam age class of 1 would have their liveweight observations multiplied by 1.02.

18
19 *Quadratic*

20
21 LWT was adjusted for dam age using the following formula;

$$22 \quad Y_a = Y_o / ((I + Lx + Qx^2) / I)$$

23 Where: Y_a = Adjusted weight

24 Y_o = Observed weight

25 x = Age of dam - standard age of dam

26 I = Intercept value

27 L = Linear component

28 Q = Quadratic component

29 The standard age of dam used in this equation was the age at which dams were at maximum
30 production.

31
32 ***Combined adjustment of LWT for individual age and age of dam***

33
34 It is common practice to adjust LWT for both individual age and age of dam. Four additional
35 adjusted observations were created by adjusting LWT for age by within group regression or age

1 intercept and then adjusting this weight for dam age using dam age class or quadratic
2 adjustment.

3 4 ***Adjustment of FATD and EMD for LWT at measurement***

5 6 *Regression*

7
8 The regression based adjustments of FATD and EMD to a standard LWT were achieved using
9 the same method as described in the regression adjustment of LWT for age.

10 11 *Quadratic*

12
13 Adjustment of carcass traits for LWT using the quadratic method was performed in the same
14 manner as described for the quadratic adjustment of LWT for age of dam.

15 16 ***Estimation of adjustment factors***

17
18 The standard regression coefficient was the average regression coefficient of LWT on age from
19 all groups in the data. For each CG a regression coefficient was calculated as described earlier
20 which weighted the observed coefficient by the number of animals in each CG. This coefficient
21 for each group was used to adjust the LWT of each animal in that group to the standard age.
22 The standard age used was the average age of all the animals in the CG.

23
24 The age intercept and the coefficients for the intercept, linear and quadratic components for
25 adjustment of liveweight for dam age were estimated using an animal model in ASREML
26 (Gilmour *et al.* 1999). The model included adjusted age (i.e. observed age minus standard
27 age), adjusted dam age (i.e. observed minus the standard age of dam), and adjusted dam age
28 squared as covariates, the fixed effect of CG (defined as flock, sex, year of measurement,
29 management group, birth type and rearing type) and the random effects of direct and maternal
30 genetic components, permanent environment due to dam, and residual. The age-intercept was
31 calculated by back-solving the linear regression equation based on the solution estimates from
32 the ASREML analysis. This analysis was then re-run after removing the dam age covariates
33 and replacing them with a dam age class. Solutions for dam age class from this analysis were
34 used to calculate the adjustment factors. Multiplicative factors were computed from the ratio of
35 dam age class 2 to dam age classes 1 and 3 respectively.

1
2 Adjustment factors for within group regression adjustment of carcass traits for LWT were
3 calculated in the same manner as described for the regression based adjustment of LWT for age.
4

5 Coefficients for the intercept, linear and quadratic components to adjust carcass traits for LWT
6 were estimated using an animal model in ASREML (Gilmour *et al.* 1999). The model included
7 adjusted weight (i.e. observed weight minus the standard weight) with linear and quadratic
8 components, the fixed effect of CG (defined as flock, sex, year of measurement, management
9 group, birth type, rear type and dam age class) and the random effects of additive genetic and
10 residual components.

11 12 ***Statistical Analysis***

13
14 Variance components were estimated for each observed and adjusted trait using an animal
15 model in a univariate analysis with ASREML (Gilmour *et al.* 1999). CG was the only fixed
16 effect fitted. CG was defined using flock, sex, year of measurement, management group, birth
17 type and rearing type. All possible combinations of adjustment methods for each trait were
18 explored to yield a total of 15 models (Table 1).
19

20 When LWT was not adjusted for dam age, dam age class was also included in the definition
21 of CG (models 1, 2 and 3 in Table 1). The analysis of LWT included the random effects of
22 direct genetic, maternal genetic, direct-maternal genetic covariance and permanent
23 environment due to dam. The analysis of FATD and EMD included the random effect of
24 animal.
25

26 [Please insert Table 1 near here]
27

28 To further investigate the differences between models 1, 6, 7, 8 and 9 an additional analysis was
29 performed. A random 10% of the animals in the data file were removed. An ASREML analysis
30 was then performed using the remaining 90% of the data to produce EBVs. LWT observations
31 from previously removed animals were regressed against the EBVs of their sires. This
32 regression analysis included the fixed effect of contemporary group in the model using the
33 GLM procedure of SAS (1990). This analysis was repeated 20 times to produce the mean,
34 minimum and maximum regression coefficients for each model.
35

1 The phenotypic relationships between observed traits, adjusted traits and the adjustment
2 variables were compared using simple and residual correlations. The simple correlations were
3 estimated over all data without fitting any fixed effects and using the CORR procedure of SAS
4 (1990). The phenotypic correlations were estimated using the GLM procedure of SAS. For
5 LWT the model included the fixed effect of CG and covariates of age and dam age. The model
6 for FATD and EMD included the fixed effect of CG and covariate for LWT with linear and
7 quadratic components.

8 9 **Results**

10
11 Descriptive statistics for the observed traits are presented in Table 2. With the exception of
12 EMD, positively skewed distributions were observed for all traits. Values in Table 2 are similar
13 to post-weaning observations presently recorded in the LAMBPLAN database for the White
14 Suffolk breed.

15
16 [Please insert Table 2 near here]

17
18 Estimates of the adjustment factors are given in Table 3. These adjustment factors were used to
19 adjust observations for each trait. Standard points in Table 3 are very similar to the means for
20 these traits given in Table 2. Within CG regression coefficients ranged from 0 to 0.87 kg/d, 0 to
21 0.16 mm/kg and 0 to 0.52 mm/kg for the regression adjustment of LWT for age, FATD for
22 LWT and EMD for LWT, respectively. Quadratic components for adjustment of LWT for dam
23 age and, FATD and EMD for LWT were all negative, indicating a decrease in performance
24 after the standard point has been reached.

25
26 [Please insert Table 3 near here]

27
28 Mean, standard deviation, minimum and maximum for the adjusted observations are given in
29 Table 4. The means of the adjusted traits are similar to those of the unadjusted traits. The
30 variation and range in the adjusted LWT were greater than those for the observed LWT for age-
31 intercept and the combined adjustment for individual age and dam age. Within group regression
32 and quadratic adjustment of FATD and EMD for weight resulted in less variation between
33 adjusted observations compared to the observed measurements. This reduction in variance was
34 more apparent for the quadratic methods of adjustment. Variation between animals in adjusted
35 LWT was greater for the age-intercept adjustment method.

1 [Please insert Table 4 near here]

2
3 Variance components estimated from the genetic analysis of the LWT traits are tabulated in
4 Table 5. Observed LWT was moderately heritable (0.27). Maternal heritabilities for LWT were
5 low (0.09 to 0.12) and so were the permanent environment effects due to dam (10 to 16% of the
6 phenotypic variance).

7
8 Both maternal heritability and permanent environment due to dam were similar across methods
9 of adjustment. While within group regression adjustment of LWT for age did not affect the
10 estimate of phenotypic variance, direct heritability was reduced. The age intercept however
11 resulted in higher phenotypic variance (27.55) and direct heritability (0.30).

12
13 [Please insert Table 5 near here]

14
15 [Please insert Figure 2 near here]

16
17 The two methods used to adjust LWT for age of dam produced adjusted observations with
18 similar phenotypic variances and heritabilities. The combined adjustment of individual age
19 using within CG regression and age of dam using both dam age class and quadratic methods
20 (models 6 and 7) produced estimates that were generally similar to each other but with lower
21 direct and slightly lower maternal heritabilities compared to the analysis of observed LWT. This
22 reduction in phenotypic variance and direct heritability was not observed when age-intercept
23 adjustment was used in combination with dam age class or quadratic adjustment methods
24 (models 8 and 9). Regression coefficients of progeny performance on sire EBV for models 8
25 and 9 were closer to 0.50, the expectation, and had a small range across replicates than those for
26 models 6 and 7.

27
28 The effect of the quadratic dam age adjustment is illustrated in Figure 2. While the quadratic
29 adjusted method is removing the majority of the dam age effect it appears that animals out of
30 very young dams are under adjusted and those out of older dams are slightly over adjusted.

31
32 Estimates of phenotypic variance, direct and maternal heritabilities of the observed and adjusted
33 carcass traits of FATD and EMD are illustrated in Table 6. Regression adjustment of FATD and
34 EMD resulted in adjusted observations with slightly lower phenotypic variance and similar

1 heritability of FATD and EMD. The quadratic adjustment slightly decreased phenotypic
2 variance and tended to increase the heritability of FATD and EMD.

3
4 [Please insert Table 6 near here]

5
6 The relationships between age, LWT and adjusted LWT are given in Table 7. A positive
7 phenotypic relationship between observed LWT and age existed ($r=0.26$). The expectation is
8 that once a trait is adjusted, its relationship with the adjustor variable will be broken. While the
9 simple correlation suggested that this relationship remained after using within group regression
10 adjustment, the phenotypic correlation suggests that the relationship between regression
11 adjusted liveweight and age was removed. Within groups age intercept adjusted LWT was not
12 significantly correlated with age ($r=0.04$, $P>0.05$) however across all data there was a
13 significant negative correlation between adjusted liveweight and age (-0.41 , $P<0.05$). Adjusted
14 LWT using both within group regression and age intercept methods were highly correlated with
15 observed LWT and each other.

16
17 [Please insert Table 7 near here]

18
19 While there were significant positive relationships between LWT and FATD and EMD ($r=0.49$
20 and 0.59 respectively, $P<0.05$) there were no significant relationships of the regression and
21 quadratic adjusted FATD and EMD with LWT ($r=-0.03$ to 0.02 , $P>0.05$). Adjusted FATD and
22 EMD using both regression and quadratic methods were highly correlated with the actual
23 observations ($r=0.79$ to 0.86 respectively, $P<0.05$).

24 25 **Discussion**

26
27 The data used in this study originated from the White Suffolk LAMBPLAN database. This
28 breed was chosen because of their good performance recording scheme, especially on the dam's
29 side. The procedures described in this study were also explored using the Poll Dorset and Texel
30 breeds (data not shown) with similar results to those reported here for the White Suffolk. It is
31 anticipated that these results would be applicable to most sheep genetic evaluation systems.

32
33 Heritability of observed liveweight, fat and eye muscle depth are similar to those previously
34 reported (Atkins *et al.* 1991; Brash *et al.* 1992; Gilmour *et al.* 1994; Fogarty 1995). With the
35 exception of age intercept and dam age adjustment of liveweight all of the adjustment methods

1 used in this study did not influence the phenotypic variance of the trait. This result is anticipated
2 and agrees with the results of Raymond (1982). There were also increases in the heritability of
3 the age intercept adjusted liveweight and quadratic adjustment of fat depth for liveweight.
4 Gregory *et al.* (1978) also observed that adjusting liveweight for age can lead to increases in
5 the heritability of post-weaning liveweight and gain. These results indicate that the adjustment
6 processes are performing as expected towards reducing systematic environmental differences
7 between animals.

8
9 While the regression adjustment of liveweight for individual age had slightly greater reductions
10 in the phenotypic variance of liveweight the heritability was lower than that of the age intercept
11 adjusted liveweight. The difference in heritability estimates indicates that the age intercept
12 method of adjustment of liveweight for individual age is preferable to the regression adjustment.
13 The latter appears to have more potential to remove genetic variation.

14
15 The regression of progeny performance on sire estimated breeding values also illustrated that
16 the age intercept produced estimated breeding values that were closer to the expectation. This
17 relationship was also more consistent across the 20 random samples of data for the age intercept
18 adjustments. The method used to adjust for age of dam did not appear to influence the
19 relationship between progeny performance and sire EBV.

20
21 Both age intercept and within group regression adjustment removed the relationship between
22 liveweight and age within groups. As anticipated there was a positive relationship between
23 within group regression adjusted liveweight and age across all data. However there was also a
24 moderate negative correlation between age intercept adjusted liveweight and age across all data.
25 This negative correlation arises from over-adjusting older animals. The cause of this over
26 adjustment is a non-linear growth curve, which was modelled as linear. The age intercept
27 adjustment method is only designed to adjust animals in a linear growth pattern. The
28 measurements used in this study were all recorded as post-weaning weights but covered a large
29 age range (110 to 320 days). Splitting the data into two age groups, firstly all animals less than
30 190 days of age and secondly all animals greater than 190 days, illustrated that animals in the
31 first data set were growing at a quicker rate than those in the second data set. This resulted in
32 the age intercept method expecting the average animals to be heavier at older ages than they
33 were actually observed. The age intercept method adjusts these older animals to be below
34 average at the standard age when in fact they were at an average weight for their age. These
35 results agree with the findings of Boggess *et al.* (1991) who concluded that linear age

1 adjustment for weaning weight at 90 days is only adequate for ages ranging up to 28 days either
2 side of weaning. Outside this range quadratic regression adjustment may be more appropriate.
3 This highlights the need to restrict the use of age intercept to groups of animals that have a
4 linear growth pattern during that period.

5
6 When dam age operated as the sole adjustment, differences between adjustment methods were
7 small. Both techniques slightly increased the phenotypic variance but did not influence the
8 heritability estimates of liveweight. Results indicate that both techniques used in this study
9 adjusted the data adequately. Differences between adjustment methods may have been
10 influenced by the fact that in this particular data set the influence of dam age was not as great as
11 commonly found in other sheep data at similar stages of growth (Gilmour 1993). The influence
12 of dam age in this study can also be gauged from the multiplicative factors used for the dam age
13 class adjustments. The differences were 1.7 to 2.4% of the liveweight of the middle dam age
14 class (Table 3). These dam age adjustments are slightly smaller than the 2 to 4 % previously
15 reported (Gilmour 1993). Dam age adjustments were also influenced by the fact that 80% of the
16 animals in the data file had dam ages within the range of 3 to 6 years. Alternatively 36, 60 and
17 4% of animals were in dam age classes 1, 2 and 3 respectively. Therefore there were insufficient
18 animals at the lower and upper bounds of dam age to produce a large dam age effect or to
19 estimate its effect with sufficient accuracy. However, when dam age adjustment methods were
20 combined with individual age adjustment the advantage of the age intercept and quadratic
21 methods (model 9) were apparent.

22
23 Quadratic adjustment of carcass traits slightly reduced the phenotypic variance but tended to
24 increase the direct heritability. The residual correlations between fat and eye muscle depth with
25 liveweight were also very small for both methods of adjustment. These results indicate that both
26 methods produce very similar adjusted observations however the heritability estimates tend to
27 indicate that the quadratic method may be slightly more appropriate for genetic evaluation
28 purposes.

29
30 When computational performance is considered, the age intercept adjustment method has the
31 additional appeal of being less demanding. Within group regression requires additional base
32 parameters to be estimated and stored for the adjustment routine. Furthermore, the actual
33 adjustment process contains an additional level of complexity, which requires 5 floating-point
34 computations for each CG. As a result the programming and running of the age intercept
35 adjustment method is substantially more efficient and less prone to programming errors.

Conclusions

Liveweight observations used in this study were successfully adjusted for individual age and dam age. Similar conclusions can be drawn when adjusting fat and eye muscle depth for liveweight. While within group regression adjustments generally had the greatest reduction in phenotypic variance they also produced the greatest reductions in heritability and had the strongest phenotypic relationships with their adjustment variables. Furthermore, within group regressions were also more computationally demanding to perform. Results indicate that age intercept and quadratic methods of adjustment for age and liveweight respectively may be more appropriate for genetic evaluation purposes. These trends are apparent in at least the three major Australian meat sheep breeds.

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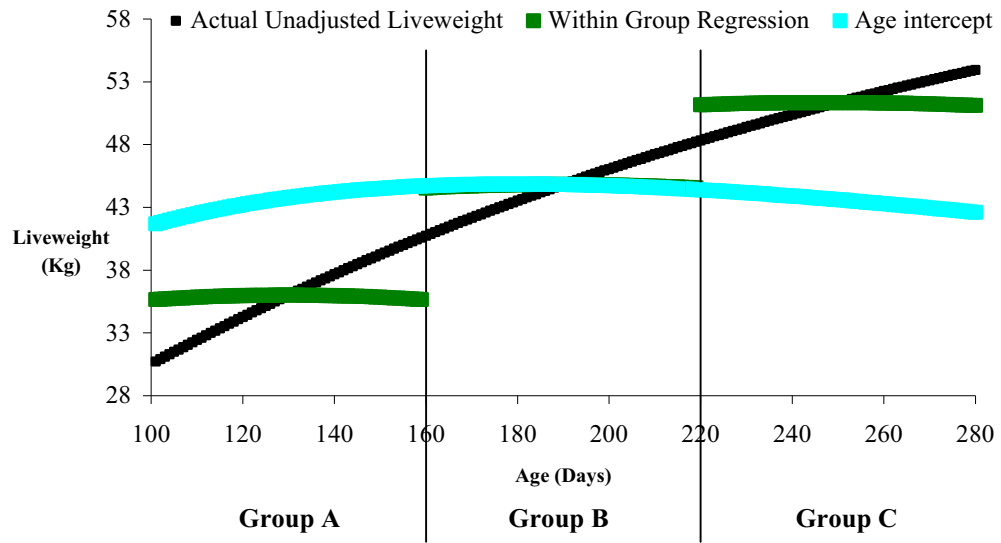
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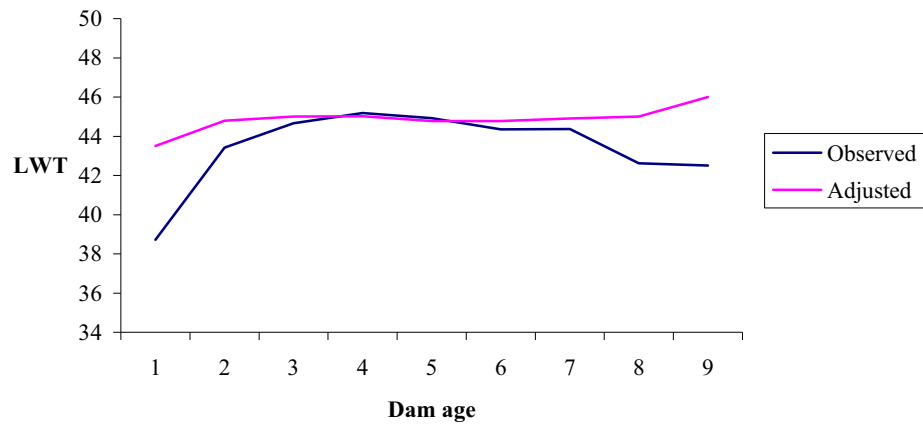
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- 1 **Figure 1 Illustration of the expected relationships between weight and age before and**
- 2 **after adjustment**



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1 **Figure 2 The observed and adjusted liveweight of sheep within dam age groups**

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Table 1 Description of the adjustment combinations for each trait analysed

Model	Dependent Variable	Adjustment for Age^a	Adjustment for Dam Age^b	Adjustment for Weight^c
1	LWT	-	-	-
2	LWT	LR	-	-
3	LWT	AI	-	-
4	LWT	-	DAC	-
5	LWT	-	QUA	-
6	LWT	LR	DAC	-
7	LWT	LR	QUA	-
8	LWT	AI	DAC	-
9	LWT	AI	QUA	-
10	FATD	-	-	-
11	FATD	-	-	LR
12	FATD	-	-	QUA
13	EMD	-	-	-
14	EMD	-	-	LR
15	EMD	-	-	QUA

LWT = post-weaning liveweight, FATD = fat depth, EMD = eye muscle depth

^{abc} LR = Linear regression, AI = Age intercept, DAC = Dam age class, QUA = Quadratic Regression

Table 2 Descriptive statistics for the observed traits (N=8418)

Trait^a	Model^b	Mean	Std Dev	Min	Max
Individual age, days		192	51	107	321
Age of dam, years		3.4	1.3	1	9
LWT, Kg	1	44.23	10.05	16.50	86.00
FATD, mm	10	2.85	1.01	0.50	7.00
EMD, mm	13	26.84	4.09	13.00	39.00

^a LWT = liveweight, FATD = fat depth, EMD = eye muscle depth

^b Refer to Table 1 for model specifications

Table 3 Estimates of the adjustment factors to standard points

Standard Points			
Standard LWT	45 kg		
Standard age	190 days		
Standard age of dam	3.5 years		
Standard CG size	50 animals		
Regression Adjustment			
	Standard	Min	Max
LWT for age	0.2845	0	0.8674
FATD for LWT	0.0666	0	0.1556
EMD for LWT	0.2911	0	0.5174
Quadratic Adjustment			
	Intercept	Linear	Quadratic
LWT for dam age	44.0992	0.4846	-0.2698
FATD for LWT	2.6936	0.0668	-0.0003
EMD for LWT	26.7249	0.2971	-0.0018
Dam Age Class Adjustment^a			
	1	2	3
Multiplicative factor	1.0237	1.00	1.0171
Age Intercept Adjustment			
X-Intercept (days)	-38.5		

^a 1 = 1 to 2 year old dams, 2 = 3 to 5 years, 3 = 6 to 10 years

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Table 4 Descriptive statistics for the adjusted variable

Trait	Model ^a	Mean	Std Dev	Min	Max
<i>LWT for age, Kg</i>					
Regression	2	44.07	10.16	10.68	86.16
Age Intercept	3	45.13	11.47	19.16	93.89
<i>LWT for dam age, Kg</i>					
Dam age classes	4	44.63	10.17	16.78	86.00
Quadratic	5	44.70	10.17	16.75	86.61
<i>LWT for age and dam age, Kg</i>					
Regression and dam age class	6	44.46	10.28	10.68	86.16
Regression and quadratic	7	44.53	10.29	10.63	86.78
Age Intercept and dam age class	8	45.53	11.60	19.50	96.04
Age Intercept and quadratic	9	45.61	11.59	19.42	94.72
<i>FATD for LWT, mm</i>					
Regression	11	2.85	0.95	0.26	7.13
Quadratic	12	2.72	0.79	0.37	6.75
<i>EMD for LWT, mm</i>					
Regression	14	26.82	3.78	14.39	39.47
Quadratic	15	26.16	2.73	15.06	38.82

^a Refer to Table 1 for model specifications

Table 5 Phenotypic variance, direct and maternal heritability estimates and mean, minimum and maximum regression of progeny performance on sire EBV for the observed and adjusted liveweight traits from the genetic analysis

Model ^a	σ_p^2	h^2	m^2	PE	dmr	Regression ^b		
						Mean	Min	Max
1	25.190	0.27	0.11	0.14	-0.66	0.57	0.26	0.82
2	24.720	0.23	0.11	0.14	-0.63			
3	27.550	0.30	0.11	0.13	-0.59			
4	28.100	0.28	0.11	0.15	-0.59			
5	27.950	0.28	0.09	0.16	-0.56			
6	25.050	0.23	0.11	0.14	-0.63	0.59	0.30	0.91
7	25.070	0.22	0.09	0.14	-0.67	0.59	0.28	0.92
8	28.020	0.29	0.11	0.13	-0.60	0.56	0.32	0.83
9	27.880	0.28	0.09	0.13	-0.55	0.55	0.31	0.80
s.e.	0.40 to 0.50	0.04 to 0.05	0.04	0.03	0.10 to 0.17			

^a Refer to Table 1 for model specifications

^b Values from 20 replicates

σ_p^2 = Phenotypic variance, h^2 = direct heritability, m^2 = maternal heritability, PE = permanent environment due to dam, dmr = direct to maternal correlation

1 **Table 6 Phenotypic variance, direct heritability and maternal heritability estimates for the**
 2 **observed and adjusted FATD and EMD traits from the genetic analysis**

Model	σ_p^2	h^2
10	0.327	0.30
11	0.317	0.30
12	0.312	0.33
s.e.	0.01	0.05
13	3.632	0.29
14	3.548	0.29
15	3.548	0.30
s.e.	0.10	0.05

4 σ_p^2 = Phenotypic variance, h^2 = direct heritability

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Table 7 Simple (above diagonal) and phenotypic (below diagonal) correlations between LWT, regression adjusted LWT, age intercept adjusted LWT and age

	Age	LWT	Regression adjusted LWT	Age intercept adjusted LWT
Age	-	0.35	0.33	-0.41
LWT	0.26	-	0.97	0.85
Regression adjusted LWT	-0.04	0.95	-	0.83
Age intercept adjusted LWT	0.04	0.95	0.97	-

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