ANIMAL BREEDING NOTES

CHAPTER 16

ANIMAL AND REDUCED ANIMAL MODELS

Animal Model (AM)

Objective: to predict the breeding value (BV) of animals based on their own records and(or) records of their relatives.

Assumptions:

- (i) Animals belong to a single population,
- (ii) Animals may have 1 or more records and covariances among records are due only to genetic factors, and
- (iii) There is either no selection in the population, or:
 - (a) If selection occurred based on records, the selection was within fixed effects, and
 - (b) If selection occurred based on the BV of animals, the relationship matrix is complete.

The AM is:

$$y = Xb + Zu + e$$

$$E[y] = Xb$$

$$var \begin{bmatrix} u \\ e \end{bmatrix} = \begin{bmatrix} G & 0 \\ 0 & R \end{bmatrix}$$

$$= \begin{bmatrix} A\sigma_A^2 & 0 \\ 0 & I\sigma_e^2 \end{bmatrix}, \text{ because of assumptions (i) and (ii).}$$

$$var(y) = ZGZ' + R,$$

$$= ZAZ'\sigma_A^2 + I\sigma_e^2$$
,

y = vector of animal records,

b = vector of unknown fixed effects,

u = vector of unknown random BV belonging to the animals making the records,

e = vector of unknown random residual effects,

X = known incidence matrix relating records to fixed effects in vector b,

Z = known incidence matrix relating records to BV in vector u.

Let

$$\alpha = \frac{\sigma_A^2}{\sigma_e^2}$$
.

Then, the mixed model equations (MME) for the **AM**, after multiplying both sides by σ_e^2 , are:

$$\begin{bmatrix} X'X & X'Z \\ Z'X & Z'Z+A^{-1}\alpha^{-1} \end{bmatrix} \begin{bmatrix} b \\ u \end{bmatrix} = \begin{bmatrix} X'y \\ Z'y \end{bmatrix}$$

where A⁻¹ is the inverse of the matrix of additive relationships among the animals with records. If animals in the pedigree of animals with records did **not** have records themselves, their BV would not be included in vector u. This would prevent the use of Henderson's rules to compute A⁻¹ directly. However, **Henderson's rules could be used if we included the BV of the animals without records in vector u**, which can be accomplished by using the following **Equivalent Animal Model (EAM)**:

$$y = Xb + \begin{bmatrix} 0 & Z \end{bmatrix} \begin{bmatrix} u_0 \\ u_1 \end{bmatrix} + e$$

$$E[y] = Xb$$

$$var \begin{bmatrix} u_0 \\ u_1 \\ e \end{bmatrix} = \begin{bmatrix} A_{00}\sigma_A^2 & A_{01}\sigma_A^2 & 0 \\ A_{10}\sigma_A^2 & A_{11}\sigma_A^2 & 0 \\ 0 & 0 & I\sigma_e^2 \end{bmatrix}$$

 $u_1 = u \text{ of the } AM,$

 u_0 = random vector of the BV of animals without records which are relatives of the animals with records.

The variance of y is:

$$var(y) = \begin{bmatrix} 0 & Z \end{bmatrix} \begin{bmatrix} A_{00} & A_{01} \\ A_{10} & A_{11} \end{bmatrix} \begin{bmatrix} 0 \\ Z' \end{bmatrix} \sigma_A^2 + I \sigma_e^2$$
$$= Z A_{11} Z' \sigma_A^2 + I \sigma_e^2$$

Thus, the E[y] and the var(y) of the AM and the EAM are the same, proving that they are equivalent models.

The MME of the **EAM** are:

$$\begin{bmatrix} X'X & 0 & X'Z \\ 0 & A^{00}\alpha^{-1} & A^{01}\alpha^{-1} \\ Z'X & A^{10}\alpha^{-1} & Z'Z+A^{11}\alpha^{-1} \end{bmatrix} \begin{bmatrix} b \\ u_0 \\ u_1 \end{bmatrix} = \begin{bmatrix} X'y \\ 0 \\ Z'y \end{bmatrix}$$

where

$$\begin{bmatrix} A^{00} & A^{01} \\ A^{10} & A^{11} \end{bmatrix} = \begin{bmatrix} A_{00} & A_{01} \\ A_{10} & A_{11} \end{bmatrix}^{-1}, \text{ computed using Henderson's rules}.$$

Remarks:

(1) Absorption of the equations for u_0 into b and u_1 yields:

$$\begin{bmatrix} X'X & X'Z \\ Z'X & Z'Z + (A^{11} - A^{10}(A^{00})^{-1}A^{01})\alpha^{-1} \end{bmatrix} \begin{bmatrix} b \\ u_1 \end{bmatrix} = \begin{bmatrix} X'y \\ Z'y \end{bmatrix}$$

$$(A^{11} - A^{10} (A^{00})^{-1} A^{01}) \alpha^{-1} = A_{11}^{-1} \alpha^{-1}$$

= $A^{-1} \alpha^{-1}$

Proof:

$$\begin{bmatrix} A^{00} & A^{01} \\ A^{10} & A^{11} \end{bmatrix} \begin{bmatrix} A_{00} & A_{01} \\ A_{10} & A_{11} \end{bmatrix} = \begin{bmatrix} I & 0 \\ 0 & I \end{bmatrix}$$

$$A^{00} A_{01} + A^{01} A_{11} = 0 ag{1}$$

$$A^{10} A_{01} + A^{11} A_{11} = I$$
 [2]

From [1]:

$$A_{01} = -(A^{00})^{-1} A^{01} A_{11}$$
 [3]

Substituting [3] for A_{01} in [2] yields:

$$-A^{10} (A^{00})^{-1} A^{01} A_{11} + A^{11} A_{11} = I$$

$$(A^{11} - A^{10} (A^{00})^{-1} A^{01}) A_{11} = I$$

$$\Rightarrow$$
 $A_{11}^{-1} = (A^{11} - A^{10} (A^{00})^{-1} A^{01})$

$$\rightarrow$$
 K'b° ϵ AM = K'b° ϵ EAM for estimable K'

- $\Rightarrow \qquad \hat{u} \text{ from AM} \qquad \equiv \quad \hat{u}_1 \text{ from EAM}$
- (2) From the equations for u_0 of the MME for the **EAM**,

$$A^{00} \alpha^{-1} \hat{u}_{0} = -A^{01} \alpha^{-1} \hat{u}_{1}$$

$$\hat{u}_{0} = -(A^{00})^{-1} A^{01} \hat{u}_{1}$$

or

$$\hat{\mathbf{u}}_0 = -(\mathbf{A}^{00})^{-1} \mathbf{A}^{01} \hat{\mathbf{u}}$$

 $\hat{\mathbf{u}}_1 = BLUP \text{ of } \mathbf{u}_1 (= \mathbf{u}) \text{ from the } \mathbf{EAM}, \text{ and }$

 $\hat{u} = BLUP \text{ of } u (= u_1) \text{ from the } AM.$

Thus, \hat{u}_0 is the BLUP of u_0 , because u_0 is a linear combination of $\hat{u}_1 = \hat{u}$, the BLUP of u.

Also, notice that:

$$\hat{\mathbf{u}}_0 = -(\mathbf{A}^{00})^{-1} \mathbf{A}^{01} \hat{\mathbf{u}}_1$$

$$= \mathbf{A}_{01} \mathbf{A}_{11}^{-1} \hat{\mathbf{u}}_1$$

Proof:

Equation [1] above is:

$$A^{00} A_{01} + A^{01} A_{11} = 0$$

$$A^{01} = -A^{00} A_{01} A$$

Substituting $-A^{00}A_{01}$ A for A^{01} in $\hat{u}_0 = -(A^{00})^{-1}$ A^{01} \hat{u}_1 yields:

$$\hat{u}_0 = -(A^{00})^{-1} (-A^{00}) A_{01} A_{11}^{-1} \hat{u}_1$$

$$\hat{\mathbf{u}}_0 = \mathbf{A}_{01} \, \mathbf{A}_{11}^{-1} \, \hat{\mathbf{u}}_1$$

or

$$\hat{u}_0 = A_{01} A_{11}^{-1} \hat{u}$$

- (3) This method used to obtain the BLUP of u₀ is actually a **general method to predict the BV of** animals not represented in vector u, but correlated to some of the elements of u (see Henderson, 1977).
- (4) Advantages of the **EAM**:
 - (a) all additive genetic relationships are used, and

(b) the MME are easy to construct.

(5) Disadvantages of the **EAM**:

- (a) there are usually more equations (i.e., more unknowns) than there are records, and
- (b) sometimes the MME do not behave well in iterative solutions (probably as a consequence of disadvantage 5a).

Reduced Animal Model (RAM)

Objective: same as the **AM** (or the **EAM**), i.e., to predict the BV of animals with and without records. However, the set of equations used to compute the BV will require less number of computations. The reduction in computations is achieved by exploiting the structure of A.

Assumptions: same as the **AM**.

Derivation of the RAM

Denote the **EAM**:

$$y = X b + \begin{bmatrix} 0 & Z \end{bmatrix} \begin{bmatrix} u_0 \\ u_1 \end{bmatrix} + e$$
 [4]

$$y \sim (X b, Z A_{11} Z' \sigma_A^2 + I \sigma_e^2)$$

as:

$$y = Xb + \dot{Z}\dot{u} + e$$

$$y \sim \left(Xb, \, \dot{Z} A_{11} \dot{Z}' \sigma_A^2 + I \sigma_e^2\right)$$
[5]

where

$$\dot{Z} = \begin{bmatrix} 0 & Z \end{bmatrix}$$

$$\dot{\mathbf{u}} = \begin{bmatrix} \mathbf{u}_0 \\ \mathbf{u}_1 \end{bmatrix}$$

Consider ordering the y and the <u>u</u> vectors of the **EAM** as follows:

$$y = \begin{bmatrix} y_p \\ y_n \end{bmatrix}$$
, and $\dot{u} = \begin{bmatrix} u_p \\ u_n \end{bmatrix}$,

where

 y_p = subvector of y containing records of animals that have progeny with records,

 y_n = subvector of y with records of animals that are nonparents,

 $u_p = \text{subvector of } \dot{u} \text{ representing the BV of parents with or without records of their own, and}$

 $u_n = \text{subvector of } \dot{u} \text{ holding the BV of nonparents.}$

The **EAM** corresponding this partitioning of the y and $\dot{\mathbf{u}}$ is:

$$\begin{bmatrix} y_{p} \\ y_{n} \end{bmatrix} = \begin{bmatrix} X_{p} \\ X_{n} \end{bmatrix} b + \begin{bmatrix} Z_{p} & 0 \\ 0 & Z_{n} \end{bmatrix} \begin{bmatrix} u_{p} \\ u_{n} \end{bmatrix} + \begin{bmatrix} e_{p} \\ e_{n} \end{bmatrix}$$
 [6]

$$E \begin{bmatrix} y_p \\ y_n \end{bmatrix} = \begin{bmatrix} X_p \\ X_n \end{bmatrix} b$$

$$\Rightarrow \qquad \text{var} \begin{bmatrix} y_{p} \\ y_{n} \end{bmatrix} = \begin{bmatrix} Z_{p} & 0 \\ 0 & Z_{n} \end{bmatrix} \begin{bmatrix} A_{pp} & A_{pn} \\ A_{np} & A_{nn} \end{bmatrix} \begin{bmatrix} Z_{p}' & 0 \\ 0 & Z_{n} \end{bmatrix} \sigma_{A}^{2} + \begin{bmatrix} I_{p} & 0 \\ 0 & I_{n} \end{bmatrix} \sigma_{e}^{2}$$

$$= ZAZ'\sigma_A^2 + I\sigma_e^2$$
$$= var(y)$$

→ the **EAM** [5] and [6] are equivalent models.

The MME for the **EAM** [6] are:

$$\begin{bmatrix} X_{p}'X_{p} + X_{n}'X_{n} & X_{p}'Z_{p} & X_{n}'Z_{n} \\ Z_{p}'X_{p} & Z_{p}'Z_{p} + A^{pp}\alpha^{-1} & -A^{pn}\alpha^{-1} \\ Z_{n}'X_{n} & -A^{np}\alpha^{-1} & Z_{n}'Z_{n} + A^{nn}\alpha^{-1} \end{bmatrix} \begin{bmatrix} b \\ u_{p} \\ u_{n} \end{bmatrix} = \begin{bmatrix} X_{p}'y_{p} + X_{n}'y_{n} \\ Z_{p}'y_{p} \\ Z_{n}'y_{n} \end{bmatrix}$$
[7]

where

$$\begin{bmatrix} A^{pp} & A^{pn} \\ A^{np} & A^{nn} \end{bmatrix} = \begin{bmatrix} A_{pp} & A_{pn} \\ A_{np} & A_{nn} \end{bmatrix}^{-1} = A^{-1}.$$

However,

$$A^{-1} = (I - \frac{1}{2} P') D^{-1} (I - \frac{1}{2} P)$$

where

$$P = \begin{bmatrix} P_{pp} & 0 \\ P_{np} & P_{nn} \end{bmatrix}, D^{-1} = \begin{bmatrix} D_p^{-1} & 0 \\ 0 & D_n^{-1} \end{bmatrix}, I = \begin{bmatrix} I_p & 0 \\ 0 & I_n \end{bmatrix},$$

But $P_{nn} = 0$, because animals in u_n have no progeny.

Thus,

$$P = \begin{bmatrix} P_{pp} & 0 \\ P_{np} & 0 \end{bmatrix}$$

and

$$A^{-1} = \begin{bmatrix} \left(I - \frac{1}{2} P_{pp}\right)^{2} & -\frac{1}{2} P_{np} \\ 0 & I_{n} \end{bmatrix} \begin{bmatrix} D_{p}^{-1} & 0 \\ 0 & D_{n}^{-1} \end{bmatrix} \begin{bmatrix} \left(I_{p} - \frac{1}{2} P_{pp}\right) & 0 \\ -\frac{1}{2} P_{np} & I_{n} \end{bmatrix}$$

$$= \begin{bmatrix} \left(I_{p}^{-1/2} P_{pp}^{-1}\right) D_{p}^{-1} \left(I_{p}^{-1/2} P_{pp}\right) + \left(\frac{1}{2} P_{np}^{-1}\right) D_{n}^{-1} \left(\frac{1}{2} P_{np}\right) & -\left(\frac{1}{2} P_{np}^{-1}\right) D_{n}^{-1} \\ -D_{n}^{-1} \left(\frac{1}{2} P_{np}\right) & D_{n}^{-1} \end{bmatrix}$$

$$= \begin{bmatrix} A_{pp}^{-1} + \frac{1}{4} P_{np}^{-1} D_{n}^{-1} P_{np} & -\frac{1}{2} P_{np}^{-1} D_{n}^{-1} \\ -\frac{1}{2} D_{n}^{-1} P_{np} & D_{n}^{-1} \end{bmatrix}$$

$$= \begin{bmatrix} A^{pp} & A^{pn} \\ A^{np} & A^{nn} \end{bmatrix}$$

$$[8]$$

Substituting [8] for the submatrices of the A⁻¹ of MME [7] yields:

$$\begin{bmatrix} X_{p}, X_{p} + X_{n}, X_{n} & X_{p}, Z_{p} & X_{n}, Z_{n} \\ Z_{p}, X_{p}, Z_{p}, Z_{p} + A_{pp}, \alpha^{-1} + \frac{1}{4} P_{np}, D_{n}^{-1} P_{np}, \alpha^{-1} & -\frac{1}{2} P_{np}, D_{n}^{-1} \alpha^{-1} \\ Z_{n}, X_{n}, & -\frac{1}{2} D_{n}^{-1} P_{np}, \alpha^{-1} & Z_{n}, Z_{n} + D_{n}^{-1}, \alpha^{-1} \end{bmatrix} \begin{bmatrix} b \\ u_{p} \\ u_{n} \end{bmatrix} = \begin{bmatrix} X_{p}, y_{p} + X_{n}, y_{n} \\ Z_{p}, y_{p} \\ Z_{n}, y_{n} \end{bmatrix} [9]$$

Remarks:

(a) The BLUP of u_n , i.e., \hat{u}_n , can be easily computed based on b° and \hat{u}_p , i.e., from the third equation of [9],

$$\begin{split} \hat{u}_n &= (Z_n{}'Z_n + D_n{}^{-1}\alpha^{-1})^{-1} \left(Z_n{}'y_n - Z_n{}'X_nb^\circ - {}^1\!\!/_2\alpha^{-1}D_n{}^{-1}P_{np}\,\hat{u}_p \right) \\ \\ \hat{u}_n &= (Z_n{}'Z_n + D_n{}^{-1}\alpha^{-1})^{-1} \left(Z_n{}'y_n - Z_n{}'X_nb^\circ \right) \\ \\ &+ (Z_n{}'Z_n + D_n{}^{-1}\alpha^{-1})^{-1} (-{}^1\!\!/_2\alpha^{-1}D_n{}^{-1}P_{nn}\,\hat{u}_n) \\ \end{split} \right\} \text{ pedigree}$$

The BLUP of the BV of the ith nonparent is:

$$\hat{u}_{n_{i}} = \left(n_{\bullet i} + d_{ii}^{\text{-1}} \, \alpha^{\text{-1}} \,\right)^{\text{-1}} \left(\begin{array}{c} y_{\bullet i \bullet} - \sum_{k} \, n_{ki} \, b_{k} \,^{\circ} \end{array} \right) + \\ \left(\begin{array}{c} n_{\bullet i} + d_{ii}^{\text{-1}} \, \alpha^{\text{-1}} \, \right)^{\text{-1}} \left[\begin{array}{c} -1/2 \, \alpha^{\text{-1}} \, d_{ii}^{\text{-1}} \, \left(\, \delta_{s_{i}} \, \hat{u}_{s_{i}} \, + \delta_{d_{i}} \, \hat{u}_{d_{i}} \, \right) \right]$$

where δ_{s_i} and δ_{d_i} are Kronecker deltas,

and

$$\begin{split} \left(n_{\bullet i} + d_{ii}^{\text{-}1} \alpha^{\text{-}1}\right)^{\text{-}1} \left(\alpha^{\text{-}1} d_{ii}^{\text{-}1}\right) & = & \frac{\left(\frac{\alpha^{\text{-}1}}{d_{ii}}\right)}{\left(\frac{n_{\bullet i} d_{ii} + \alpha^{\text{-}1}}{d_{ii}}\right)} \\ & = & \frac{\alpha^{\text{-}1}}{n_{\bullet i} d_{ii} + \alpha^{\text{-}1}} \end{split}$$

Thus,

$$\hat{u}_{n_{i}} = \frac{d_{ii}}{n_{\bullet i} \, d_{ii} + \alpha^{\text{-}1}} \Bigg| \, y_{\bullet i \bullet} - \sum_{k} n_{ki} \, b_{k}^{\circ} \, \Bigg| + \frac{\alpha^{\text{-}1}}{n_{\bullet i} \, d_{ii} + \alpha^{\text{-}1}} \, \left[\, - \frac{1}{2} \Big(\, \delta_{s_{i}} \, \hat{u}_{s_{i}} + \delta_{d_{i}} \, \hat{u}_{d_{i}} \, \Big) \, \right]$$

If nonparents have only one record each, $n_{\bullet i} = 1$, then

$$\hat{u}_{n_{i}} = \frac{d_{ii}}{d_{ii} + \alpha^{-1}} \left[y_{i} - b_{i} \circ \right] + \frac{\alpha^{-1}}{d_{ii} + \alpha^{-1}} \left[-\frac{1}{2} \left(\delta_{s_{i}} \hat{u}_{s_{i}} + \delta_{d_{i}} \hat{u}_{d_{i}} \right) \right]$$

Let

$$w_n^{ii} = \frac{d_{ii}}{d_{ii} + \alpha^{-1}}$$

Note that,

$$1 - w_n^{ii} = \left(\frac{d_{ii} + \alpha^{-1}}{d_{ii} + \alpha^{-1}}\right) - \left(\frac{d_{ii}}{d_{ii} + \alpha^{-1}}\right)$$
$$= \frac{\alpha^{-1}}{d_{ii} + \alpha^{-1}}$$

Thus,

$$\hat{\mathbf{u}}_{n_{i}} = \mathbf{w}_{n}^{ii} \left(\mathbf{y}_{i} - \mathbf{b}_{i}^{\circ} \right) + \left(1 - \mathbf{w}_{n}^{ii} \right) \left[-\frac{1}{2} \left(\delta_{s_{i}} \, \hat{\mathbf{u}}_{s_{i}} + \delta_{d_{i}} \, \hat{\mathbf{u}}_{d_{i}} \right) \right]$$

(b) The matrix $Z_n'Z_n + D_n^{-1} \alpha^{-1}$ is diagonal. Thus, the equations for u_n can be readily absorbed into b and u_p (Hint: obtain $(\alpha Z_n D_n Z_n' + I)^{-1}$ in all equations). The absorption of u_n into b and u_p is as follows:

(i)
$$X_{p}'X_{p} + X_{n}'X_{n} - X_{n}'Z_{n}(Z_{n}'Z_{n} + D_{n}^{-1}\alpha^{-1})^{-1}Z_{n}'X_{n}$$

$$= X_{p}'X_{p} + X_{n}'[I - Z_{n}(Z_{n}'Z_{n} + D_{n}^{-1}\alpha^{-1})^{-1}Z_{n}']X_{n}$$

$$= X_{p}'X_{p} + X_{n}'(\alpha Z_{n}D_{n}Z_{n}' + I)^{-1}X_{n}$$

(ii)
$$X_{p}'Z_{p} - X_{n}'Z_{n}(Z_{n}'Z_{n} + D_{n}^{-1}\alpha^{-1})^{-1}(-1/2D_{n}^{-1}P_{np}\alpha^{-1})$$

$$= X_{p}'Z_{p} + X_{n}'Z_{n}(Z_{n}'Z_{n} + D_{n}^{-1}\alpha^{-1})^{-1}(-D_{n}^{-1}\alpha^{-1} + Z_{n}'Z_{n} - Z_{n}'Z_{n})(1/2P_{np})$$

$$= X_{p}'Z_{p} + X_{n}'Z_{n}[I - (Z_{n}'Z_{n} + D_{n}^{-1}\alpha^{-1})^{-1}Z_{n}'Z_{n}](1/2P_{np})$$

$$= X_{p}'Z_{p} + X_{n}'[I - Z_{n}(Z_{n}'Z_{n} + D_{n}^{-1}\alpha^{-1})^{-1}Z_{n}'](1/2Z_{n}P_{np})$$

$$= X_{n}'Z_{p} + X_{n}'(\alpha Z_{n}D_{n}Z_{n}' + 1)^{-1}(1/2Z_{n}P_{np})$$

$$\begin{split} (iii) & Z_p{'}Z_p + A\alpha^{-1} + (1/2P_{np}{'})(D_n{}^{-1}\alpha^{-1})(1/2P_{np}) \\ & - (1/2P_{np}{'})(D_n{}^{-1}\alpha^{-1})(Z_n{'}Z_n + D_n{}^{-1}\alpha^{-1})^{-1}(D_n{}^{-1}\alpha^{-1})(1/2P_{np}) \\ & = Z_p{'}Z_p + A\alpha^{-1} + (1/2P_{np}{'})(D_n{}^{-1}\alpha^{-1})[D_n\alpha - (ZZ_n + D_n{}^{-1}\alpha^{-1})^{-1}](D_n{}^{-1}\alpha^{-1})(1/2P_{np}) \\ & = Z_p{'}Z_p + A\alpha^{-1} + (1/2P_{np}{'})(D_n{}^{-1}\alpha^{-1}) \\ & + Z_n{'}Z_n - Z_n{'}Z_n)[D_n\alpha - (Z_n{'}Z_n + D_n{}^{-1}\alpha^{-1})^{-1}](D_n{}^{-1}\alpha^{-1})(1/2P_{np}) \\ & = Z_p{'}Z_p + A\alpha^{-1} + (1/2P_{np}{'})[I - I + Z_n{'}Z_n(Z_n{'}Z_n + D_n{}^{-1}\alpha^{-1})^{-1}] \\ & (D_n{}^{-1}\alpha^{-1} + Z_n{'}Z_n - Z_n{'}Z_n)(1/2P_{np}) \\ & = Z_p{'}Z_p + A\alpha^{-1} + (1/2P_{np}{'})[Z_n{'}Z_n - Z_n{'}Z_n(Z_n{'}Z_n + D_n{}^{-1}\alpha^{-1})^{-1}Z_n{'}Z_n](1/2P_{np}) \\ & = Z_p{'}Z_p + A\alpha^{-1} + (1/2P_{np}{'}Z_n{'})[I - Z_n{'}Z_n{'}Z_n + D_n{}^{-1}\alpha^{-1})^{-1}Z_n{'}](1/2Z_nP_{np}) \end{split}$$

$$= Z_p'Z_p + A\alpha^{-1} + (\frac{1}{2}P_{np}'Z_n')(\alpha Z_n D_n Z_n' + I)^{-1}(\frac{1}{2}Z_n P_{np})$$

$$\begin{split} (iv) \qquad & X_p{'}y_p + X_n{'}y_n - X_n{'}Z_n(Z_n{'}Z_n + D_n{^{-1}}\alpha^{-1})^{-1}Z_n{'}y_n \\ \\ & = & X_p{'}y_p + X_n{'}[I - Z_n(Z_n{'}Z_n + D_n{^{-1}}\alpha^{-1})^{-1}Z_n{'}]y_n \\ \\ & = & X_p{'}y_p + X_n{'}(\alpha Z_nD_nZ_n{'} + I)^{-1}y_n \end{split}$$

$$\begin{split} (v) \qquad & Z_{p}{'}y_{p} - (1/2P_{np}{'})(D_{n}{^{-1}}\alpha^{-1})(Z_{n}{'}Z_{n} + D_{n}{^{-1}}\alpha^{-1})^{-1}Z_{n}{'}y_{n} \\ \\ & = \quad Z_{p}{'}y_{p} - (1/2P_{np}{'})(-Z_{n}{'}Z_{n} + Z_{n}{'}Z_{n} + D_{n}{^{-1}}\alpha^{-1})(Z_{n}{'}Z_{n} + D_{n}{^{-1}}\alpha^{-1})^{-1}Z_{n}{'}y_{n} \\ \\ & = \quad Z_{p}{'}y_{p} + (1/2P_{np}{'})[Z_{n}{'}Z_{n}(Z_{n}{'}Z_{n} + D_{n}{^{-1}}\alpha^{-1})^{-1}Z_{n}{'} + Z_{n}{'}]y_{n} \\ \\ & = \quad Z_{p}{'}y_{p} + (1/2P_{np}{'}Z_{n}{'})[I + Z_{n}(Z_{n}{'}Z_{n} + D_{n}{^{-1}}\alpha^{-1})^{-1}Z_{n}{'}]y_{n} \\ \\ & = \quad Z_{p}{'}y_{p} + (1/2P_{np}{'}Z_{n}{'})(\alpha Z_{n}D_{n}Z_{n}{'} + I)^{-1}y_{n} \end{split}$$

Let $\mathbf{R}_2 = (\alpha \mathbf{Z}_n \mathbf{D}_n \mathbf{Z}_n' + \mathbf{I})$

Then, the MME after absorbing u_n are:

$$\begin{bmatrix} X_{p}, X_{p} + X_{n}, X_{2}^{-1} X_{n} & X_{p}, Z_{p} + X_{n}, X_{2}^{-1} Z_{n} P_{np}(1/2) \\ Z_{p}, X_{p} + (1/2) P_{np}, Z_{n}, X_{2}^{-1} X_{n} & Z_{p}, Z_{p} + A_{pp}^{-1} \alpha^{-1} + (1/2) P_{np}, Z_{n}, X_{2}^{-1} Z_{n} P_{np}(1/2) \end{bmatrix} \begin{bmatrix} b^{\circ} \\ \hat{u}_{p} \end{bmatrix} = \begin{bmatrix} X_{p}, Y_{p} + X_{n}, X_{2}^{-1} Y_{n} \\ Z_{p}, Y_{p} + (1/2) P_{np}, Z_{n}, X_{2}^{-1} Y_{n} \end{bmatrix} \begin{bmatrix} 10 \end{bmatrix}$$

The MME [10] are those for the model:

$$\begin{bmatrix} y_{p} \\ y_{n} \end{bmatrix} = \begin{bmatrix} X_{p} \\ X_{n} \end{bmatrix} b + \begin{bmatrix} Z_{p} \\ \frac{1}{2} Z_{n} P_{np} \end{bmatrix} u_{p} + \begin{bmatrix} e_{p} \\ Z_{n} \phi_{n} + e_{n} \end{bmatrix}$$
[11]

$$E \begin{bmatrix} y_p \\ y_n \end{bmatrix} = \begin{bmatrix} X_p \\ X_n \end{bmatrix} b$$

$$\text{var} \begin{bmatrix} u_p \\ ----- \\ e_p \\ Z_n \phi_n + e_n \end{bmatrix} = \begin{bmatrix} A_{pp} \alpha & | & 0 & & 0 \\ ---- & | & ---- \\ 0 & | & I & & 0 \\ 0 & | & 0 & Z_n D_n Z_n' \alpha + I \end{bmatrix} \sigma_e^2$$

$$\Rightarrow \qquad var \left[\begin{array}{c} y_p \\ y_n \end{array} \right] = \left[\begin{array}{c} Z_p \\ \frac{1}{2} \, Z_n \, P_{np} \end{array} \right] A_{pp} \alpha \left[Z_p \right] \quad \frac{1}{2} \, P_{np} \, Z_n \, d_p^2 + \left[\begin{array}{cc} I & 0 \\ 0 & Z_n \, D_n \, Z_n \, \alpha + I \end{array} \right] \, \sigma_e^2$$

The **EAM** [11] is called the Reduced Animal Model (RAM).

The **RAM** can also be derived starting from the EAM [6] by expressing the BV of the nonparents in terms of the BV and the Mendelian sampling of their parents, i.e., let:

$$u_n = \frac{1}{2} P_{np} u_p + \varphi_n$$
 [12]

where

 P_{np} = incidence matrix relating nonparents to their known parents,

 φ_n = vector of Mendelian sampling terms for nonparents, where

Substituting u_n in the second set of equations of [6] for expression [12] yields the EAM:

$$\begin{bmatrix} y_{p} \\ y_{n} \end{bmatrix} = \begin{bmatrix} X_{p} \\ X_{n} \end{bmatrix} b + \begin{bmatrix} Z_{p} & 0 \\ \frac{1}{2} Z_{n} P_{np} & Z_{n} \end{bmatrix} \begin{bmatrix} u_{p} \\ \phi_{n} \end{bmatrix} + \begin{bmatrix} e_{p} \\ e_{n} \end{bmatrix}$$
[13]

$$E\begin{bmatrix} y_p \\ y_n \end{bmatrix} = \begin{bmatrix} X_p \\ X_n \end{bmatrix} b$$

$$\text{var} \begin{bmatrix} u_p \\ \phi_n \\ --- \\ e_p \\ e_n \end{bmatrix} = \begin{bmatrix} A_{pp}\alpha & 0 & \mid & 0 & 0 \\ 0 & D_n\alpha & \mid & 0 & 0 \\ ---- & --- & \mid & --- & -- \\ 0 & 0 & \mid & I_p & 0 \\ 0 & 0 & \mid & 0 & I_n \end{bmatrix} \sigma_e^2$$

$$\operatorname{var} \left[\begin{array}{c} y_{p} \\ y_{n} \end{array} \right] = \left[\begin{array}{cc} Z_{p} & 0 \\ \frac{1}{2} Z_{n} P_{np} & Z_{n} \end{array} \right] \left[\begin{array}{cc} A_{pp} \alpha & 0 \\ 0 & D_{n} \alpha \end{array} \right] \left[\begin{array}{cc} Z_{p} & \frac{1}{2} P_{np} Z_{n} \\ 0 & Z_{n} \end{array} \right] \sigma_{e}^{2} + \left[\begin{array}{cc} I_{p} & 0 \\ 0 & I_{n} \end{array} \right] \sigma_{e}^{2}$$

The MME for the **EAM** [13] are:

$$\begin{bmatrix} X_{p}, X_{p} + X_{n}, X_{n} & X_{p}, Z_{p} + \frac{1}{2}X_{n}, Z_{n}P_{np} & X_{n}, Z_{n} \\ Z_{p}, X_{p} + \frac{1}{2}P_{np}, Z_{n}, X_{n} & Z_{p}, Z_{p} + A_{pp}^{-1}\alpha^{-1} + \frac{1}{4}P_{np}, Z_{n}, Z_{n}P_{np} & \frac{1}{2}P_{np}, Z_{n}, Z_{n} \\ Z_{n}, X_{n} & \frac{1}{2}Z_{n}, Z_{n}P_{np} & Z_{n}, Z_{n} + D_{n}^{-1}\alpha^{-1} \end{bmatrix} \begin{bmatrix} b^{\circ} \\ \hat{u}_{p} \\ \phi_{n} \end{bmatrix} = \begin{bmatrix} X_{p}, Y_{p} + X_{n}, Y_{n} \\ Z_{p}, Y_{p} + \frac{1}{2}P_{np}, Z_{n}, Y_{n} \\ Z_{n}, Y_{n} \end{bmatrix} \begin{bmatrix} 14 \end{bmatrix}$$

Notice that ϕ_n is uncorrelated to u_p , thus, it can be placed together with e_n to form a new residual, \dot{e}_n , where

$$\dot{\mathbf{e}}_{\mathbf{n}} = \mathbf{Z}_{\mathbf{n}} \mathbf{\phi}_{\mathbf{n}} + \mathbf{e}_{\mathbf{n}}$$

The resulting **EAM** is the RAM shown in equation [11] above.

Remarks:

(a) If nonparents have **several records** for a trait, the $var(\dot{e}) = var(Z_n\phi_n + e_n)$ is block-diagonal. For instance, if the i^{th} animal has two records:

$$\begin{aligned} var \begin{bmatrix} \dot{e}_{i1} \\ \dot{e}_{i2} \end{bmatrix} &= var \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} [\phi_i] + \begin{bmatrix} e_{i1} \\ e_{i2} \end{bmatrix} \right\} \\ &= \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} [d_{ii}\alpha][1\ 1] + \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right\} \sigma_e^2 \\ &= \left\{ \begin{bmatrix} d_{ii}\alpha + 1 & d_{ii}\alpha \\ d_{ii}\alpha & d_{ii}\alpha + 1 \end{bmatrix} \right\} \sigma_e^2 \\ &= \begin{bmatrix} d_{ii}\sigma_A^2 + \sigma_e^2 & d_{ii}\sigma_A^2 \\ d_{ii}\sigma_A^2 & d_{ii}\sigma_A^2 + \sigma_e^2 \end{bmatrix}$$

If nonparents have only 1 record, then

$$var(e) = var(\varphi_n + e_n)$$

= diagonal matrix

Thus, for the ith animal with 1 record

var(
$$\dot{e}$$
) = $(d_{ii} \alpha + 1) \sigma_e^2$
= $(d_{ii} \sigma_A^2 + \sigma_e^2)$

(b) The number of equations for the RAM will be fewer than for the other EAM which include either u_n or ϕ_n , thus they are solved faster than the other EAM. If the BLUP of some nonparents were wanted, they could be backsolved using the formulae developed in remark (a) for the MME in equation [9]. Also, backsolutions can be obtained using the MME in equation [14] for the EAM [13]. From the equations for ϕ_n , the BLUP of ϕ_n is:

$$\begin{split} \hat{\phi}_n &= (Z_n{}'Z_n + D^1\alpha^{-1})^{-1}[Z_n{}'y_n - Z_n{}'X_nb^\circ - {}^1\!\!/_2 Z_n{}'Z_nP_{np}\,\hat{u}_p] \\ \\ \hat{\phi}_n &= (Z_n{}'Z_n + D^1\alpha^{-1})^{-1}[Z_n{}'y_n - Z_n{}'X_nb^\circ] \\ \\ &+ (Z_n{}'Z_n + D^1\alpha^{-1})^{-1}({}^1\!\!/_2 Z_n{}'Z_nP_{np}\,\hat{u}_p] \end{split} \} \text{ data} \\ \\ &+ (Z_n{}'Z_n + D^1\alpha^{-1})^{-1}({}^1\!\!/_2 Z_n{}'Z_nP_{np}\,\hat{u}_p] \end{split}$$

From equation [12], the BLUP of u_n is:

$$\hat{\mathbf{u}}_{\mathrm{n}} = \frac{1}{2} P_{\mathrm{np}} \hat{\mathbf{u}}_{\mathrm{p}} + \hat{\varphi}_{\mathrm{n}}$$

For the i^{th} animal with $n_{\bullet i}$ records:

$$\hat{\phi}_{n_i} = \frac{1}{n_{\bullet i} + \frac{\alpha^{-1}}{d_{ii}}} \left[y_{\bullet i \bullet} - \sum_k n_{ki} b_k \circ \right] - \frac{n_{\bullet i}}{n_{\bullet i} + \frac{\alpha^{-1}}{d_{ii}}} \left[\frac{1}{2} \left(\delta_{s_i} \hat{u}_{s_i} + \delta_{d_i} \hat{u}_{d_i} \right) \right]$$

$$\hat{\phi}_{n_{i}} = \frac{d_{ii}}{n_{\bullet i} d_{ii} + \alpha^{-1}} \left[y_{\bullet i \bullet} - \sum_{k} n_{ki} b_{k} \circ \right] - \frac{n_{\bullet 1} d_{ii}}{n_{\bullet 1} d_{ii} + \alpha^{-1}} \left[\frac{1}{2} \left(\delta_{s_{i}} \hat{u}_{s_{i}} + \delta_{d_{i}} \hat{u}_{d_{i}} \right) \right]$$

and

$$\hat{u}_{n_i} = \frac{1}{2} \left(\delta_{s_i} \, \hat{u}_{s_i} + \delta_{d_i} \, \hat{u}_{d_i} \right) + \hat{\phi}_{n_i}$$

Example for the AM and the RAM

Animal	Sex	Weaning Weight (kg)	Sire	Dam	Mgs
1	M				
2	F		1		
3	M	292	1	2	1
4	M	286	1		
5	M	304	1		
6	F	256	3	2	1
7	F	261	3	6	3
8	F	266	4	7	3
9	F	270	5	8	4
10	F	275	5	9	5
11	M	289	3	6	3
12	M	285	4	7	3
13	F	265	4	8	4
14	M	290	5	9	5
15	F	288	5	10	5

Assumptions

$$\sigma_{A}^{2} = 22 \text{ kg}^{2}, \ \sigma_{e}^{2} = 88 \text{ kg}^{2}, \ \sigma_{P}^{2} = 110 \text{ kg}^{2}$$

$$\Rightarrow h^{2} = \frac{\sigma_{A}^{2}}{\sigma_{P}^{2}} = 0.2; \ \alpha = \frac{\sigma_{A}^{2}}{\sigma_{e}^{2}} = 0.25 \ \Rightarrow \ \alpha^{-1} = 4.0$$

[A] Consider

$$\begin{array}{rcl} y_{ijk} &=& \mu + sex_i + animal_j + residual_{ijk} \\ E[y_{ijk}] &=& \mu + sex_i \\ var(y_{ijk}) &=& var\left(animal_j\right) + var\left(residual_{ijk}\right) \\ &=& a_{jj}\,\sigma_A^2 + \sigma_e^2 \\ cov(y_{ijk},y_{i'j'k'}) &=& cov(animal_j,\,animal_{j'}) + cov(residual_{ijk},\,residual_{i'j'k'}) \\ &=& a_{jj'}\,\sigma_A^2 + \delta\,\sigma_e^2,\,where \\ \delta &=& \begin{cases} 1 & \text{if } ijk = i'\,j'\,k' & \left(\text{diagonal element}\right) \\ 0 & \text{otherwise} & \left(\text{offdiagonal element}\right) \end{cases} \end{array}$$

In matrix notation, the **AM** is:

$$y = Xb + Zu + e$$

$$E[y] = Xb$$

$$var\begin{bmatrix} u \\ e \end{bmatrix} = \begin{bmatrix} G & 0 \\ 0 & R \end{bmatrix}$$

$$= \begin{bmatrix} A\sigma_A^2 & 0 \\ 0 & I\sigma_e^2 \end{bmatrix}$$

$$= \begin{bmatrix} A\alpha & 0 \\ 0 & I \end{bmatrix} \sigma_e^2$$

$$= \begin{bmatrix} A(0.25) & 0 \\ 0 & I \end{bmatrix} (88)$$

Explicitly, the vectors and matrices of the **AM** model are:

$$\begin{bmatrix} 292 \\ 286 \\ 10 \\ 304 \\ 10 \\ 256 \\ 01 \\ 261 \\ 270 \\ = \begin{bmatrix} 0 & 1 \\ 0 &$$

and

$$G = \operatorname{cov} \left\{ \begin{bmatrix} u_3 \\ u_4 \\ \vdots \\ u_{15} \end{bmatrix}, [u_3 \ u_4 \cdots u_{15}] \right\}.$$

However, to build A^{-1} (and A) directly using Henderson's rules, we need base animals 1 and 2. Thus, instead of **AM**, we will use the following **EAM**:

$$y = Xb + [0 \quad Z] \begin{bmatrix} u_o \\ u_r \end{bmatrix} + e$$

$$u_0 = [u_1 u_2]'$$

$$u_r = [u_3 u_4 ... u_{15}]'$$

 $0 = 13 \times 2$ submatrix of zeroes for u_0

Thus,

$$E[y] = Xb$$

and

$$\operatorname{var} \begin{bmatrix} u_{0} \\ u_{r} \\ e \end{bmatrix} = \begin{bmatrix} A_{00}(0.25) & A_{0r}(0.25) & 0 \\ A_{r0}(0.25) & A_{rr}(0.25) & 0 \\ 0 & 0 & I \end{bmatrix} (88)$$

The inverse of the relationship matrix A is:

$$A^{-1} = (I - \frac{1}{2} P') D^{-1} (I - \frac{1}{2} P)$$

The diagonal elements of A and the elements of D and D^{-1} are:

i	diagonal of A	diagonal of D	diagonal of D ⁻¹
1	1.0	1.0	1.0
2	1.0	0.75	1.3333
3	1.25	0.5	2.0
4	1.0	0.75	1.3333
5	1.0	0.75	1.3333
6	1.375	0.4375	2.2857
7	1.5	0.34375	2.9091
8	1.171875	0.375	2.6667
9	1.1484375	0.45703125	2.1880
10	1.32421875	0.462890625	2.1603
11	1.5	0.34375	2.9091
12	1.171875	0.375	2.6667
13	1.3359375	0.45703125	2.1880
14	1.32421875	0.462890625	2.1603
15	1.412109375	0.4189453125	2.3869

The matrix $(I - \frac{1}{2}P)$ is:

<u> </u>														7
-1/2	1													
-1/2	$-\frac{1}{2}$	1												
-1/2	0	0	1											
-1/2	0	0	0	1										
0	$-\frac{1}{2}$	$-\frac{1}{2}$	0	0	1									
0	0	$-\frac{1}{2}$	0	0	$-\frac{1}{2}$	1								
0	0	0	$-\frac{1}{2}$	0	0	$-\frac{1}{2}$	1							
0	0	0	0	$-\frac{1}{2}$	0	0	$-\frac{1}{2}$	1						
0	0	0	0	$-\frac{1}{2}$	0	0	0	$-\frac{1}{2}$	1					
0	0	$-\frac{1}{2}$	0	0	$-\frac{1}{2}$	0	0	0	0	1				
0	0	0	$-\frac{1}{2}$	0	0	$-\frac{1}{2}$	0	0	0	0	1			
0	0	0	$-\frac{1}{2}$	0	0	0	$-\frac{1}{2}$	0	0	0	0	1		
0	0	0	0	$-\frac{1}{2}$	0	0	0	$-\frac{1}{2}$	0	0	0	0	1	
0	0	0	0	$-\frac{1}{2}$	0	0	0	0	$-\frac{1}{2}$	0	0	0	0	1

The A⁻¹ matrix, built by Henderson's rules, is:

2.5	-0.167	-1.0	-0.667	-0.667	0	0	0	0	0	0	0	0	0	0
	2.405	-0.429	0	0	-1.143	0	0	0	0	0	0	0	0	0
		4.026	0	0	0.312	-1.455	0	0	0	-1.455	0	0	0	0
			3.214	0	0	1.333	-0.786	0	0	0	-1.333	-1.094	0	0
				3.557	0	0	0.547	-0.0138	-0.483	0	0	0	-1.080	-1.194
					3.740	-1.455	0	0	0	-1.455	0	0	0	0
						4.242	-1.333	0	0	0	-1.333	0	0	0
							3.761	-1.094	0	0	0	-1.094	0	0
								3.268	-1.080	0	0	0	-1.080	0
		Symmetric							2.757	0	0	0	0	-1.194
										2.909	0	0	0	0
											2.667	0	0	0
												2.188	0	0
													2.160	0
														2.387

The MME for the **EAM** are:

Γ	6	0	0	0	1	1	1	0	0	0	0	0	1	1	0	1	0	b_1	[1746
		7	0	0	0	0	0	1	1	1	1	1	0	0	1	0	1	b ₂		1881
			10	-0.667	-4	-2.667	_2.667	0	0	0	0	0	0	0	0	0	0	u ₁		0
				9.619	-1.714	0	0	-4.572	0	0	0	0	0	0	0	0	0	u ₂		0
					17.104	0	0	1.247	-5.818	0	0	0	-5.818	0	0	0	0	u ₃		292
						13.855	0	0	5.333	-3.145	0	0	0	-5.333	-4.376	0	0	u ₄		286
							15.229	0	0	2.188	-0.0554	-1.933	0	0	0	-4.321	-4.774	u ₅		304
								15.961	-5.818	0	0	0	-5.818	0	0	0	0	u ₆	=	256
									17.97	-5.333	0	0	0	-5.3330	0	0	0	u ₇		261
										16.043	-4.375	0	0	0	-4.376	0	0	u_8		266
											14.073	-4.321	0	0	0	-4.321	0	u ₉		270
						Symmetric						12.028	0	0	0	0	-4.774	u ₁₀		275
													12.636	0	0	0	0	u_{11}		289
														11.667	0	0	0	u ₁₂		285
															9.752	0	0	u ₁₃		265
																9.641	0	u ₁₄		290
																	10.548	u ₁₅ _		_ 288]

The vector of solutions for the **EAM** is:

$\left[\begin{array}{c c}b_1^{\circ}\end{array}\right]$		291.6375
b ₂ °		269.0400
$\hat{\mathbf{u}}_1$		- 0.4410
$\hat{\mathbf{u}}_2$		- 2.0758
û ₃		- 2.1405
û ₄		-1.5212
û ₅		3.5970
û ₆	=	-3.5008
û ₇		-3.3839
$\hat{\mathrm{u}}_{8}$		- 2.4499
û ₉		0.9258
$\hat{\mathbf{u}}_{10}$		3.3714
$\hat{\mathbf{u}}_{11}$		- 2.8061
$\hat{\mathbf{u}}_{12}$		-2.8113
û ₁₃		- 2.1962
$\hat{\mathbf{u}}_{14}$		1.8570
$\begin{bmatrix} \hat{\mathbf{u}}_{15} \end{bmatrix}$		4.9514 _

[16-25]

The MME for the AM (i.e., the EAM with u_0 absorbed into b and u_1) are:

Γ	6	0	1	1	1	0	0	0	0	0	1	1	0	1	[0	$\begin{bmatrix} b_1 \end{bmatrix}$	[1746]
		7	0	0	0	1	1	1	1	1	0	0	1	0	1	b_2	1881
			15.094	-1.103	-1.103	0.301	-5.818	0	0	0	-5.818	0	0	0	0	u ₃	292
				13.140	-0.714	-0.085	5.333	-3.145	0	0	0	-5.33	-4.3760	0	0	u ₄	286
					14.515	-0.085	0	2.188	-0.055	-1.934	0	0	0	-4.321	- 4.774	u ₅	304
						13.788	-5.818	0	0	0	-5.818	0	0	0	0	u ₆	256
							17.97	-5.333	0	0	0	-5.333	0	0	0	u ₇	= 261
								16.043	-4.376	0	0	0	-4.376	0	0	u ₈	266
									14.073	-4.321	0	0	0	-4.321	0	u ₉	270
										12.028	0	0	0	0	- 4.774	u ₁₀	275
											12.636	0	0	0	0	u ₁₁	289
						Symmetric						11.667	0	0	0	u ₁₂	285
													9.752	0	0	u ₁₃	265
														9.641	0	u ₁₄	290
															10.548	$\begin{bmatrix} u_{15} \end{bmatrix}$	288

The solution vector for the MME for the $\mbox{\bf AM}$ (i.e., the $\mbox{\bf EAM}$ with u_0 absorbed) is:

b_1°		291.6375
b ₂ °		269.040
û ₃		-2.1405
û ₄		-1.5212
û ₅		3.5970
$\hat{\mathrm{u}}_{6}$		-3.5008
û ₇	_	-3.3839
$\hat{\mathrm{u}}_{8}$	_	- 2.4499
û ₉		0.9258
$\hat{\mathbf{u}}_{10}$		3.3714
$\hat{\mathbf{u}}_{11}$		-2.8061
$\hat{\mathbf{u}}_{12}$		-2.8113
û ₁₃		-2.1962
$\hat{\mathbf{u}}_{14}$		1.8570
$\hat{\mathbf{u}}_{15}$		4.9514

The BLUP of u_0 based on the solution vector of the AM are:

$$\hat{u}_0 = A_{0r} A_{rr}^{-1} \hat{u}_1$$

$$\hat{\mathbf{u}}_0 = -(\mathbf{A}^{00})^{-1} \mathbf{A}^{0r} \hat{\mathbf{u}}_1$$

[B] The **RAM** is:

$$\begin{bmatrix} y_p \\ y_n \end{bmatrix} = \begin{bmatrix} X_p \\ X_n \end{bmatrix} b + \begin{bmatrix} Z_p \\ \frac{1}{2} I_n P_{np} \end{bmatrix} u_p + \begin{bmatrix} e_p \\ I_n \phi_n + e_n \end{bmatrix}$$

$$E \begin{bmatrix} y_p \\ y_n \end{bmatrix} = \begin{bmatrix} X_p \\ X_n \end{bmatrix} b$$

$$\operatorname{var}\begin{bmatrix} u_{p} \\ e_{p} \\ \phi_{n} + e_{n} \end{bmatrix} = \begin{bmatrix} A_{pp}(\frac{1}{4}) & | & 0 & 0 \\ - - - & | & - - & - - - \\ 0 & | & I & 0 \\ 0 & | & 0 & D_{n}(\frac{1}{4}) + I \end{bmatrix} (88)$$

The diagonals of A_p , D_p and D^{-1} are:

p	a_{pp}	d_{pp}	d_{pp}^{-1}
1	1.0	1.0	1.0
2	1.0	0.75	1.3333
3	1.25	0.5	2.0
4	1.0	0.75	1.3333
5	1.0	0.75	1.3333
6	1.375	0.4375	2.2857
7	1.5	0.34375	2.9091
8	1.171875	0.375	2.6667
9	1.1484375	0.45703125	2.1880
10	1.32421875	0.462890625	2.1603

The A_{pp}^{-1} matrix is:

The diagonal elements of D_n and D^{-1} are:

n	d_{nn}	d_{nn}^{-1}
11	0.34375	2.9091
12	0.375	2.6667
13	0.45703125	2.1880
14	0.462890625	2.1603
15	0.4189453125	2.3869

The matrix P_{np}, which relates parents to progeny, is:

$$P_{np} = \begin{bmatrix} 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

Thus, the **RAM** looks like:

$$\begin{bmatrix} y_{p} \\ y_{n} \end{bmatrix} = \begin{bmatrix} X_{p} \\ X_{n} \end{bmatrix} b + \begin{bmatrix} Z_{p} \\ \frac{1}{2} P_{np} \end{bmatrix} u_{p} + \begin{bmatrix} e_{p} \\ \phi_{n} + e_{n} \end{bmatrix}$$

The matrix $R_2 = D_n (1/4) + I$ is:

$$R_2 = \begin{bmatrix} 1.0859375 \\ & 1.09375 \\ & & 1.11425781 \\ & & & 1.11572266 \\ & & & & 1.10473633 \end{bmatrix}$$

The MME for the **RAM** are:

$$\begin{bmatrix} 5.731 & 0 & | & 0 & 0 & 1.460 & 1.457 & 1.448 & 0.460 & 0.457 & 0 & 0.448 & 0 \\ 6.803 & | & 0 & 0 & 0 & 0.449 & 0.4533 & 1.0 & 1.0 & 1.449 & 1.0 & 1.453 \\ | & 10.0 & -0.667 & -4.0 & -2.667 & -2.667 & 0 & 0 & 0 & 0 & 0 \\ | & & 9.619 & -1.714 & 0 & 0 & -4.571 & 0 & 0 & 0 & 0 \\ | & & & 14.425 & 0 & 0 & -1.432 & -5.818 & 0 & 0 & 0 \\ | & & & & 9.453 & 0 & 0 & 2.895 & -5.109 & 0 & 0 \\ | & & & & & 11.32 & 0 & 0 & 2.188 & -1.992 & -4.094 \\ | & & & & & & 13.282 & -5.818 & 0 & 0 & 0 \\ | & & & & & & 15.532 & -5.333 & 0 & 0 \\ | & & & & & & & 14.079 & -4.376 & 0 \\ | & & & & & & & & 12.137 & -4.321 \\ | & & & & & & & & 9.868 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ 1826.511 \\ ---- \\ u_1 \\ u_2 \\ u_3 \\ u_4 \\ u_5 \\ u_6 \\ u_{10} \\ u_{$$

The vector of solutions for the **RAM** is:

The vector of deviations of the BV of the nonparents from the midparental BV, i.e., the vector ϕ_n , is:

$$\begin{split} \phi_n &= & (I + D_n^{-1} \; \alpha^{-1})^{-1} \; [y_n - X_n b_n - 1/2 \; P_{np} \, u_p] \\ \\ \phi_n &= & W^{-1} [y_n - X_n b_n - 1/2 \; P_{np} \, u_p] \end{split}$$

where

$$\begin{array}{lcl} W^{-1} & = & diag \; \{w_a^{\; ii}\} \\ \\ W^{-1} & = & diag \; \left\{ \frac{d_{n_{ii}}}{d_{n_{ii}} + \alpha^{-1}} \right\} \end{array}$$

$$W^{-1} = diag \begin{cases} 0.34375 * (0.34375 + 4)^{-1} \\ 0.375 * (0.375 + 4)^{-1} \\ 0.45703125 * (0.45703125 + 4)^{-1} \\ 0.472890625 * (0.462890625 + 4)^{-1} \\ 0.4189453125 * (0.4189453125 + 4)^{-1} \end{cases}$$

The BLUP of φ_n for the i^{th} nonparent is:

$$\widehat{\phi}_{n_i} = w_n^{ii} \left[y_{n_i} - b_{n_i}^o - \frac{1}{2} \big(\widehat{u}_{s_i} + \widehat{u}_{d_i} \big) \right]$$

Thus,

$$\begin{vmatrix} \hat{\phi}_{11} \\ \hat{\phi}_{12} \\ \hat{\phi}_{13} \\ \hat{\phi}_{14} \\ \hat{\phi}_{0} \end{vmatrix} = \begin{bmatrix} 0.0791[\ 289 - 291.6375 - \frac{1}{2}(-2.1405 - 3.5008)] \\ 0.0857[\ 285 - 281.6375 - \frac{1}{2}(-1.5212 - 3.3839)] \\ 0.1025[\ 265 - 269.0403 - \frac{1}{2}(-1.5212 - 2.4499)] \\ 0.1037[\ 290 - 291.6375 - \frac{1}{2}(3.5970 + 0.9258)] \\ 0.0948[\ 288 - 269.0403 - \frac{1}{2}(3.5970 + 3.3714)] \end{bmatrix}$$

$$= \begin{bmatrix} 0.0144939 \\ -0.3586502 \\ -0.2106119 \\ -0.4043159 \\ 1.4670774 \end{bmatrix}$$

and

$$\hat{\boldsymbol{u}}_{n_i} \quad = \quad {}^{1\!\!}/_{\!2} \left(\, \hat{\boldsymbol{u}}_{s_i} + \hat{\boldsymbol{u}}_{d_i} \, \right) + \hat{\boldsymbol{\varphi}}_{n_i}$$

$$\begin{bmatrix} \hat{u}_{11} \\ \hat{u}_{12} \\ \hat{u}_{13} \\ \hat{u}_{14} \\ \hat{u}_{15} \end{bmatrix} = \begin{bmatrix} -2.82065 + 0.01449 \\ -2.45255 - 0.35865 \\ -1.98555 - 0.26061 \\ 2.26140 - 0.40432 \\ 3.48420 + 1.45708 \end{bmatrix} = \begin{bmatrix} -2.8061 \\ -2.8113 \\ -2.1962 \\ 1.8570 \\ 4.9514 \end{bmatrix}.$$

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