Inverse Numerator Relationship Matrix

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Structure of A^{-1}

▶ Look at a simple example of A and A^{-1}

Table 1: Pedigree Used To Compute Inverse Numerator Relationship Matrix

Calf	Sire	Dam
1	NA	NA
2	NA	NA
3	NA	NA
4	1	2
5	3	2

Numerator Relationship Matrix A

$$A = \begin{bmatrix} 1.0000 & 0.0000 & 0.0000 & 0.5000 & 0.0000 \\ 0.0000 & 1.0000 & 0.0000 & 0.5000 & 0.5000 \\ 0.0000 & 0.0000 & 1.0000 & 0.0000 & 0.5000 \\ 0.5000 & 0.5000 & 0.0000 & 1.0000 & 0.2500 \\ 0.0000 & 0.5000 & 0.5000 & 0.2500 & 1.0000 \end{bmatrix}$$

Inverse Numerator Relationship Matrix A^{-1}

$$A^{-1} = \begin{bmatrix} 1.5000 & 0.5000 & 0.0000 & -1.0000 & 0.0000 \\ 0.5000 & 2.0000 & 0.5000 & -1.0000 & -1.0000 \\ 0.0000 & 0.5000 & 1.5000 & 0.0000 & -1.0000 \\ -1.0000 & -1.0000 & 0.0000 & 2.0000 & 0.0000 \\ 0.0000 & -1.0000 & -1.0000 & 0.0000 & 2.0000 \end{bmatrix}$$

Conclusions

- $ightharpoonup A^{-1}$ has simpler structure than A itself
- Non-zero elements only at positions of parent-progeny and parent-mate positions
- ▶ Parent-mate positions are positive, parent-progeny are negative

Henderson's Rules

▶ Based on LDL-decomposition of *A*

$$A = L * D * L^T$$

where *L* Lower triangular matrix *D* Diagonal matrix

- ► Why?
 - ▶ matrices L and D can be inverted directly, we 'll see how . . .
 - construct $A^{-1} = (L^T)^{-1} * D^{-1} * L^{-1}$

Example

$$L = \begin{bmatrix} 1.0 & 0.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 1.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 1.0 & 0.0 & 0.0 \\ 0.5 & 0.5 & 0.0 & 1.0 & 0.0 \\ 0.0 & 0.5 & 0.5 & 0.0 & 1.0 \end{bmatrix}$$

$$D = \begin{bmatrix} 1.0 & 0.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 1.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 1.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 0.0 & 0.5 & 0.0 \\ 0.0 & 0.0 & 0.0 & 0.0 & 0.5 \end{bmatrix}$$

 \rightarrow Verify that $A = L * D * L^T$

Decomposition of True Breeding Value

▶ True breeding value (u_i) of animal i

$$u_i = \frac{1}{2}u_s + \frac{1}{2}u_d + m_i$$

▶ Do that for all animals in pedigree

Decomposition for Example

$$u_1 = m_1$$

$$u_2 = m_2$$

$$u_3 = m_3$$

$$u_4 = \frac{1}{2}u_1 + \frac{1}{2}u_2 + m_4$$

$$u_5 = \frac{1}{2}u_3 + \frac{1}{2}u_2 + m_5$$

Matrix Vector Notation

- Define vectors a and m as
- \triangleright Coefficients of u_s and u_d into matrix P

$$a = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \end{bmatrix}, m = \begin{bmatrix} m_1 \\ m_2 \\ m_3 \\ m_4 \\ m_5 \end{bmatrix}, P = \begin{bmatrix} 0.0 & 0.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 0.0 & 0.0 & 0.0 \\ 0.5 & 0.5 & 0.0 & 0.0 & 0.0 \\ 0.0 & 0.5 & 0.5 & 0.0 & 0.0 \end{bmatrix}$$

Result: Decomposition of true breeding values

$$a = P \cdot a + m$$

Decomposition of Variance

▶ Analogous decomposition of $var(u_i)$

$$\begin{aligned} var(u_i) &= var(1/2u_s + 1/2u_d + m_i) \\ &= var(1/2u_s) + var(1/2u_d) + \frac{1}{2} * cov(u_s, u_d) + var(m_i) \\ &= 1/4var(u_s) + 1/4var(u_d) + \frac{1}{2} * cov(u_s, u_d) + var(m_i) \end{aligned}$$

► From the definition of *A*

$$egin{aligned} ext{var}(u_i) &= (1+F_i)\sigma_u^2 \ ext{var}(u_s) &= (1+F_s)\sigma_u^2 \ ext{var}(u_d) &= (1+F_d)\sigma_u^2 \ ext{cov}(u_s,u_d) &= (A)_{sd}\sigma_u^2 = 2F_i\sigma_u^2 \end{aligned}$$

Variance of Mendelian Sampling Terms

- ▶ What is $var(m_i)$?
- ▶ Solve equation for $var(u_i)$ for $var(m_i)$

$$var(m_i) = var(u_i) - 1/4var(u_s) - 1/4var(u_d) - 2*cov(u_s, u_d)$$

Insert definitions from A

$$var(m_i) = (1 + F_i)\sigma_u^2 - 1/4(1 + F_s)\sigma_u^2 - 1/4(1 + F_d)\sigma_u^2 - \frac{1}{2} * 2 * F_i\sigma_u^2$$
$$= \left(\frac{1}{2} - \frac{1}{4}(F_s + F_d)\right)\sigma_u^2$$

▶ True, for both parents s and d of animal i are known

Unknown Parents

Only parent s of animal i is known

$$\begin{aligned} u_i &= \frac{1}{2}u_s + m_i \\ var(m_i) &= \left(1 - \frac{1}{4}(1 + F_s)\right)\sigma_u^2 \\ &= \left(\frac{3}{4} - \frac{1}{4}F_s\right)\sigma_u^2 \end{aligned}$$

Both parents are unknown

$$u_i = m_i$$
 $var(m_i) = \sigma_u^2$

Recursive Decomposition

▶ True breeding values of s and d can be decomposed into

$$u_{s} = \frac{1}{2}u_{ss} + \frac{1}{2}u_{ds} + m_{s}$$
$$u_{d} = \frac{1}{2}u_{sd} + \frac{1}{2}u_{dd} + m_{d}$$

where ss sire of s ds dam of s sd sire of d dd dam of d

Example

▶ Add animal 6 with parents 4 and 5 to our example pedigree

Calf	Sire	Dam
1	NA	NA
2	NA	NA
3	NA	NA
4	1	2
5	3	2
6	4	5

First Step Of Decomposition

$$u_{1} = m_{1}$$

$$u_{2} = m_{2}$$

$$u_{3} = m_{3}$$

$$u_{4} = \frac{1}{2}u_{1} + \frac{1}{2}u_{2} + m_{4}$$

$$u_{5} = \frac{1}{2}u_{3} + \frac{1}{2}u_{2} + m_{5}$$

$$u_{6} = \frac{1}{2}u_{4} + \frac{1}{2}u_{5} + m_{6}$$

Decompose Parents

$$u_{1} = m_{1}$$

$$u_{2} = m_{2}$$

$$u_{3} = m_{3}$$

$$u_{4} = \frac{1}{2}m_{1} + \frac{1}{2}m_{2} + m_{4}$$

$$u_{5} = \frac{1}{2}m_{3} + \frac{1}{2}m_{2} + m_{5}$$

$$u_{6} = \frac{1}{2}\left(\frac{1}{2}(u_{1} + u_{2}) + m_{4}\right) + \frac{1}{2}\left(\frac{1}{2}(u_{3} + u_{2}) + m_{5}\right) + m_{6}$$

$$= \frac{1}{4}(u_{1} + u_{2}) + \frac{1}{2}m_{4} + \frac{1}{4}(u_{3} + u_{2}) + \frac{1}{2}m_{5} + m_{6}$$

Decompose Grand Parents

▶ Only animal 6 has true breeding values for grand parents

$$u_6 = \frac{1}{4}(u_1 + u_2) + \frac{1}{2}m_4 + \frac{1}{4}(u_3 + u_2) + \frac{1}{2}m_5 + m_6$$

$$= \frac{1}{4}m_1 + \frac{1}{4}m_2 + \frac{1}{4}m_3 + \frac{1}{4}m_2 + \frac{1}{2}m_4 + \frac{1}{2}m_5 + m_6$$

$$= \frac{1}{4}m_1 + \frac{1}{2}m_2 + \frac{1}{4}m_3 + \frac{1}{2}m_4 + \frac{1}{2}m_5 + m_6$$

Summary

$$u_{1} = m_{1}$$

$$u_{2} = m_{2}$$

$$u_{3} = m_{3}$$

$$u_{4} = \frac{1}{2}m_{1} + \frac{1}{2}m_{2} + m_{4}$$

$$u_{5} = \frac{1}{2}m_{3} + \frac{1}{2}m_{2} + m_{5}$$

$$u_{6} = \frac{1}{4}m_{1} + \frac{1}{2}m_{2} + \frac{1}{4}m_{3} + \frac{1}{2}m_{4} + \frac{1}{2}m_{5} + m_{6}$$

Matrix-Vector Notation

Use vectors a and m again

$$a = \begin{bmatrix} a_1 \\ a_2 \\ a_3 \\ a_4 \\ a_5 \\ a_6 \end{bmatrix}, \ m = \begin{bmatrix} m_1 \\ m_2 \\ m_3 \\ m_4 \\ m_5 \\ m_6 \end{bmatrix}, \ L = \begin{bmatrix} 1.00 & 0.00 & 0.00 & 0.00 & 0.00 & 0.00 \\ 0.00 & 1.00 & 0.00 & 0.00 & 0.00 & 0.00 \\ 0.00 & 0.00 & 1.00 & 0.00 & 0.00 & 0.00 \\ 0.50 & 0.50 & 0.00 & 1.00 & 0.00 & 0.00 \\ 0.00 & 0.50 & 0.50 & 0.50 & 0.00 & 1.00 & 0.00 \\ 0.25 & 0.50 & 0.25 & 0.50 & 0.50 & 1.00 \end{bmatrix}$$

Result of recursive decomposition of u_i

$$a = L \cdot m$$

Variance From Recursive Decomposition

$$var(u) = var(L \cdot m)$$

= $L \cdot var(m) \cdot L^{T}$

where var(m) is the variance-covariance matrix of all components in vector m.

- ▶ covariances of components m_i , $cov(m_i, m_i) = 0$ for $i \neq j$
- var(m_i) computed as shown before

Result

• variance-covariance matrix var(m) can be written as $D*\sigma_u^2$ where D is diagnoal

$$\to A = L \cdot D \cdot L^T$$

Inverse of A Based on L and D

- ▶ Matrix A was decomposed into $A = L \cdot D \cdot L^T$
- Get A^{-1} as $A^{-1} = (L^T)^{-1}D^{-1}L^{-1}$
- $ightharpoonup D^{-1}$ is diagonal again with elements

$$(D^{-1})_{ii} = 1/(D)_{ii}$$

Inverse of L

Compute m based on the two decompositions of u

$$u = P \cdot u + m$$
 and $u = L \cdot m$

Solve both for m and set them equal

$$m = u - P \cdot u = (I - P) \cdot u$$
 and $m = L^{-1} \cdot u$

$$(I-P)\cdot u=L^{-1}\cdot u$$

and

$$L^{-1} = I - P$$

Example

Calf	Sire	Dam
1	NA	NA
2	NA	NA
3	NA	NA
4	1	2
5	3	2

Matrix D^{-1}

▶ Because *D* is diagonal

$$D = \left[\begin{array}{cccccc} 1.0 & 0.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 1.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 1.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 0.0 & 0.5 & 0.0 \\ 0.0 & 0.0 & 0.0 & 0.0 & 0.5 \end{array} \right]$$

▶ We get D^{-1} as

$$D^{-1} = \left[\begin{array}{ccccc} 1.0 & 0.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 1.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 1.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 0.0 & 2.0 & 0.0 \\ 0.0 & 0.0 & 0.0 & 0.0 & 2.0 \end{array} \right]$$

Matrix L^{-1}

- ▶ Use $L^{-1} = I P$
- ► Matrix *P* from simple decomposition

$$P = \begin{bmatrix} 0.0 & 0.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 0.0 & 0.0 & 0.0 \\ 0.5 & 0.5 & 0.0 & 0.0 & 0.0 \\ 0.0 & 0.5 & 0.5 & 0.0 & 0.0 \end{bmatrix}$$

Therefore

$$L^{-1} = I - P = \begin{bmatrix} 1.0 & 0.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 1.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 1.0 & 0.0 & 0.0 \\ -0.5 & -0.5 & 0.0 & 1.0 & 0.0 \\ 0.0 & -0.5 & -0.5 & 0.0 & 1.0 \end{bmatrix}$$

Decomposition of A^{-1} I

$$A^{-1} = (L^{-1})^T \cdot D^{-1} \cdot L^{-1}$$

$$(L^{-1})^T \cdot D^{-1}$$

$$\begin{bmatrix} 1.0 & 0.0 & 0.0 & -0.5 & 0.0 \\ 0.0 & 1.0 & 0.0 & -0.5 & -0.5 \\ 0.0 & 0.0 & 1.0 & 0.0 & -0.5 \\ 0.0 & 0.0 & 0.0 & 1.0 & 0.0 \\ 0.0 & 0.0 & 0.0 & 0.0 & 1.0 \end{bmatrix} \cdot \begin{bmatrix} 1.0 & 0.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 1.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 1.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 0.0 & 2.0 & 0.0 \\ 0.0 & 0.0 & 0.0 & 0.0 & 2.0 \end{bmatrix}$$

$$= \begin{bmatrix} 1.0 & 0.0 & 0.0 & -1.0 & 0.0 \\ 0.0 & 1.0 & 0.0 & -1.0 & -1.0 \\ 0.0 & 0.0 & 0.0 & 0.0 & 2.0 \\ 0.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 0.0 & 0.0 \end{bmatrix}$$

Decomposition of A^{-1} II

$$A^{-1} = (L^{-1})^T \cdot D^{-1} \cdot L^{-1}$$

$$\begin{bmatrix} 1.0 & 0.0 & 0.0 & -1.0 & 0.0 \\ 0.0 & 1.0 & 0.0 & -1.0 & -1.0 \\ 0.0 & 0.0 & 1.0 & 0.0 & -1.0 \\ 0.0 & 0.0 & 0.0 & 2.0 & 0.0 \\ 0.0 & 0.0 & 0.0 & 0.0 & 2.0 \end{bmatrix} \cdot \begin{bmatrix} 1.0 & 0.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 1.0 & 0.0 & 0.0 & 0.0 \\ 0.0 & 0.0 & 1.0 & 0.0 & 0.0 \\ -0.5 & -0.5 & 0.0 & 1.0 & 0.0 \\ 0.0 & -0.5 & -0.5 & 0.0 & 1.0 \end{bmatrix}$$

$$= \begin{bmatrix} 1.5 & 0.5 & 0.0 & -1.0 & 0.0 \\ 0.5 & 2.0 & 0.5 & -1.0 & -1.0 \\ 0.0 & 0.5 & 1.5 & 0.0 & -1.0 \\ -1.0 & -1.0 & 0.0 & 2.0 & 0.0 \\ 0.0 & -1.0 & -1.0 & 0.0 & 2.0 \end{bmatrix}$$

Henderson's Rules

- ► Both Parents Known
 - ▶ add 2 to the diagonal-element (i, i)
 - ▶ add -1 to off-diagonal elements (s, i), (i, s), (d, i) and (i, d)
 - ▶ add $\frac{1}{2}$ to elements (s, s), (d, d), (s, d), (d, s)
- Only One Parent Known
 - ▶ add $\frac{4}{3}$ to diagonal-element (i, i)
 - ▶ add $-\frac{2}{3}$ to off-diagonal elements (s, i), (i, s)
 - ▶ add $\frac{1}{3}$ to element (s, s)
- Both Parents Unknown
 - \triangleright add 1 to diagonal-element (i, i)
- Valid without inbreeding