



Two blending approaches

MiX99 course on genomic prediction

COURSE DAY 1, March 9th, 2026



Contents

- Introduction
- Theory in brief
- Going through details using small example
- How to get variance components for indices
- Conclusions



What is blending?

- In this presentation, blending means integration of external information, usually from Interbull MACE, to domestic evaluation
- Several different methods exist
 - Bayesian methods, where external information is used as prior for breeding values
 - Pseudo-observation-based methods, where DRPs weighted with ERCs are included to domestic model
- The pseudo-observation-based method will be explained, with two different implementations, for models with and without covariance functions

Steps in blending

Blending approach has three steps

1. Calculation of amount of external information for selected animals
2. Calculation of pseudo-observations for animals
3. Running evaluation model with pseudo-observations

Selecting which animals to blend

- Aim is to choose animals that have additional information in external evaluation compared to domestic evaluation
- Can be chosen by comparing R^2_{dom} to R^2_{ext} , for example by requiring that
$$R^2_{\text{ext}} > 0.50 \text{ and } R^2_{\text{ext}} - R^2_{\text{dom}} > 0.01 \text{ for all traits}$$
- Thresholds for choosing animals can be modified depending on the traits

1. Calculation of amount of external information for selected animals

ERC (effective record contribution)

- **ERC** for external animals is calculated based on R^2 for domestic and external (MACE) evaluations
- Apax99 provides single and multitrait ERC calculation
- The amount of additional information in external evaluation calculated for each trait separately:

$$\mathbf{ERC}_{\text{blend}} = \mathbf{ERC}_{\text{external}} - \mathbf{ERC}_{\text{domestic}}$$

2. Calculation of pseudo-observations for animals

DRP (deregressed proofs)

- **DRP_{ext}** and **DRP_{dom}** for selected external animals is calculated based on corresponding **ERC** and **EBV**
- Pseudo-observation for external information

$$DRP_{blend} = \frac{ERC_{ext} * DRP_{ext} - ERC_{dom} * DRP_{dom}}{ERC_{ext} - ERC_{dom}}$$

- Weight for observation

$$Weight = ERC_{ext} - ERC_{dom}$$

- DRP_{blend} contains additional information in external evaluation

3. Running evaluation model with pseudo-observations

This step depends on the model and traits in question

- Data, model definition, and variance components needs to be modified
- Assumes genetic correlation of 1 between external and domestic evaluation
- Next example will illustrate necessary modifications

Example

Simple test-day model for milk and protein 1 lactation

```

MODEL
milk      = CalvAge(age age2) Lac(t1 t2 t3 t4 t5) HYS pe(t1 t2 t5|ID) g(t1 t2 t5|ID)
protein   = CalvAge(age age2) Lac(t1 t2 t3 t4 t5) HYS pe(t1 t2 t5|ID) g(t1 t2 t5|ID)

RANDOM     pe g
PEDIGREE  g am+p 0.3333

```

- The model is fitted for two data sets: Luke only, and Luke + 3 other herds data (FULL).
- Blending approach is demonstrated by blending information from FULL data set to Luke evaluation.
- Information is blended as 305d DRPs.
- Assume genetic correlation 1 between evaluations!



Data description

LUKE data

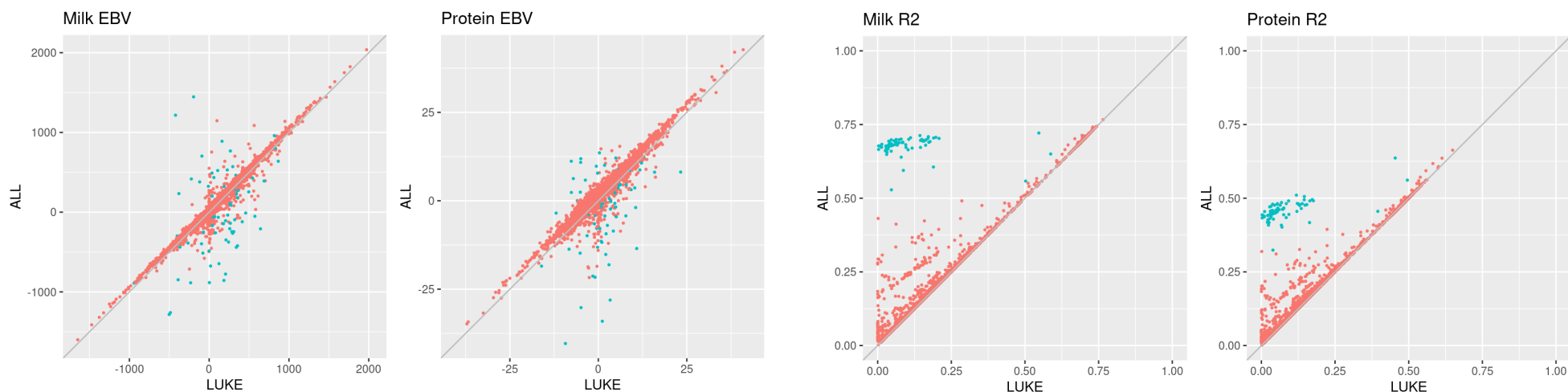
DEPENDENT VARIABLES:						
TR	TR-NAME	N-OBS	MEAN	SD	MINIMUM	MAXIMUM
1	milk	20052	27.115	4.3267	13.438	40.414
2	protein	20022	1.0156	0.13948	0.53244	1.4324
Number of Individuals.....					2751	
Number of UPG.....					1	
Number of Individuals with data records.....					537	
Number of Data Records Processed					20052	
Total Number of Accepted Observations					40074	
Number of Trait Patterns					2	
Maximum Number of Trait Patterns					3	

FULL data

DEPENDENT VARIABLES:						
TR	TR-NAME	N-OBS	MEAN	SD	MINIMUM	MAXIMUM
1	milk	22301	27.059	4.3097	13.438	40.414
2	protein	22267	1.0070	0.14149	0.53244	1.4324
Number of Individuals.....					2751	
Number of UPG.....					1	
Number of Individuals with data records.....					607	
Number of Data Records Processed					22301	
Total Number of Accepted Observations					44568	
Number of Trait Patterns					2	
Maximum Number of Trait Patterns					3	



EBVs and R2 and choosing which animals are blended



- Animals having **FULL MILK R2 – LUKE MILK R2 > 0.05** and **FULL MILK R2 > 0.50** were chosen to be blended. (Marked with different color)
- In total 70 cows and 3 bulls were chosen



Calculating ERC

```
#-----
# AccurType: Reliability method
      40 p # multitrait ERC calculations
# Reliabilities: Filename of the reliability information (IDs and r2s)
      R2forBLEND.dat
# Pedigree:
      resfarm.pedi
# Number of ERCs:
      2
# G-matrix:
      G0.par
# R-matrix:
      R0.par
# tol_ERC: tolERC MaxIter_ERC smallestERC tolNewton MaxIter_Newton
      1e-8 100 1e-10 1e-4 20
# tol_r2:  tolr2 MinIter Proportion_of_iter Num_LastIter_meet_tolr2
      1e-7 10 0.5 10
#-----
```

Separately for FULL and Luke evaluation:

- Full pedigree
- Only R2s for 73 animals!
- G0 and R0 for 305d variances

```
[lst102@sorvi-nd3 ERC]$ cat G0.par
475409.72500      9933.05886
      9933.05886      345.99609
[lst102@sorvi-nd3 ERC]$ cat R0.par
522059.61397      18225.04345
      18225.04345      672.78187
```



Calculating DRP

```

TITLE          MT-DRP calculation

DATAFILE       DRPINPUT.dat
PEDFILE        resfarm.pedi
PARFILE        DRPcalculation.para

INTEGER        ID MEAN
REAL           EBVmilk EBVprotein ERCmilk ERCprotein
MISSING        -99999.0

MODEL
EBVmilk        = MEAN  g(ID) ! WEIGHT=ERCmilk
EBVprotein     = MEAN  g(ID) ! WEIGHT=ERCprotein

RANDOM          g
PEDIGREE        g  am+p 0.3333
TMPDIR         ./tmpMiX
    
```

FULL

DEPENDENT VARIABLES:						
TR	TR-NAME	N-OBS	MEAN	SD	MINIMUM	MAXIMUM
1	EBVmilk	73	-16.530	542.65	-1283.5	1446.3
2	EBVprotein	73	-2.7486	11.358	-40.367	13.552
Number of Individuals.....					2751	
Number of UPG.....					1	
Number of Individuals with data records.....					73	
Number of Data Records Processed					73	
Total Number of Accepted Observations					146	
Number of Trait Patterns					1	
Maximum Number of Trait Patterns					3	
WEIGHTS:						
TR	TR-NAME	N-OBS	MEAN	SD	MINIMUM	MAXIMUM
1	EBVmilk	73	1.4817	0.20311	0.78985	1.8669
2	EBVprotein	73	1.4395	0.16016	0.89511	1.8076

LUKE

DEPENDENT VARIABLES:						
TR	TR-NAME	N-OBS	MEAN	SD	MINIMUM	MAXIMUM
1	EBVmilk	73	123.56	332.02	-941.52	866.18
2	EBVprotein	73	1.4660	6.5556	-16.047	23.262
Number of Individuals.....					2751	
Number of UPG.....					1	
Number of Individuals with data records.....					73	
Number of Data Records Processed					73	
Total Number of Accepted Observations					146	
Number of Trait Patterns					1	
Maximum Number of Trait Patterns					3	
WEIGHTS:						
TR	TR-NAME	N-OBS	MEAN	SD	MINIMUM	MAXIMUM
1	EBVmilk	73	0.80378E-01	0.21406	0.29958E-05	1.3083
2	EBVprotein	73	0.12271	0.30273	0.26412E-05	1.8441



Calculating pseudo-observation and weight

Separately for milk and protein:

- $DRP_{blend} = \frac{ERC_{ext} * DRP_{ext} - ERC_{dom} * DRP_{dom}}{ERC_{ext} - ERC_{dom}}$

- $Weight = ERC_{ext} - ERC_{dom}$

- For 3 animal weight for milk was negative
- observations for these animals were omitted in blending

⇒ In total 70 animals will be blended

Model allowing DRPs to be included

- Observations are test-day records, modelled by random regression curve

$$Y_{i,dim} = X\beta + \Phi_{dim}p_i + \Phi_{dim}u_i + \epsilon_{i,dim}$$

- Model DRPs with Φ_{305} :

$$DRP_i = \mu + \Phi_{305}u_i + \epsilon_{i,drp},$$

- Assume the same u_i for test-day observations and DRPs!, ie genetic correlation = 1
- Variation due to PE and residual is included to $\epsilon_{i,drp}$.

CLIM requires some changes

Original model

```
MODEL
milk      = CalvAge(age age2) Lac(t1 t2 t3 t4 t5) HYS pe(t1 t2 t5|ID) g(t1 t2 t5|ID)
protein   = CalvAge(age age2) Lac(t1 t2 t3 t4 t5) HYS pe(t1 t2 t5|ID) g(t1 t2 t5|ID)
```

Blending model

```
MODEL
milk(1)    = CalvAge(age age2) Lac(t1:5) HYS pe(t1 t2 t5|ID) g(t1 t6 t2 t6 t5 t6 |ID)@1
protein(1) = CalvAge(age age2) Lac(t1:5) HYS pe(t1 t2 t5|ID) g(t6 t1 t6 t2 t6 t5 |ID)@1
#DRPS
milk(2)    = CalvAge( - - ) Lac( D5 ) HYS pe(- - - ) g(t1 t6 t2 t6 t5 t6 |ID)@1 ! WEIGHT = mwght
protein(2) = CalvAge( - - ) Lac( D5 ) HYS pe(- - - ) g(t6 t1 t6 t2 t6 t5 |ID)@1 ! WEIGHT = pwght
```

- () after trait name refers to trait group. 1 = test-day observation, 2 = DRP
- The "@1" combines unknowns accros traits, here for genetic effect
- Covariable t6 is always zero.
- HYS for DRPs is always 1
- Covariable values for DRPs are 305d sum of DIM covariables!



Genetic effect in original model

```
MODEL
milk      = ... g(t1 t2 t5|ID)
protein   = ... g(t1 t2 t5|ID)
```

Breeding value for given DIM:

$$S_{\text{dim}}^{\text{orig}} = \begin{bmatrix} t1_{\text{dim}} & 0 & t2_{\text{dim}} & 0 & t5_{\text{dim}} & 0 \\ 0 & t1_{\text{dim}} & 0 & t2_{\text{dim}} & 0 & t5_{\text{dim}} \end{bmatrix}$$

$$EBV_{\text{dim},i}^{\text{orig}} = S_{\text{dim}}^{\text{orig}} u_i$$

"Solani"-File: Solutions for the Individual Genetic Effects

Column | Description

```
-----
1      Individual ID
2      Number of Descendants
3      Number of Observations
4      Solution for Trait 1  milk  and Factor  T1(ID)
5      Solution for Trait 2  protein and Factor  T1(ID)
6      Solution for Trait 1  milk  and Factor  T2(ID)
7      Solution for Trait 2  protein and Factor  T2(ID)
8      Solution for Trait 1  milk  and Factor  T5(ID)
9      Solution for Trait 2  protein and Factor  T5(ID)
```

Genetic effect in BLEND model

```
MODEL
milk(1)      = g(t1 t6 t2 t6 t5 t6 |ID)@1
protein(1)   = g(t6 t1 t6 t2 t6 t5 |ID)@1
#DRPS
milk(2)      = g(t1 t6 t2 t6 t5 t6 |ID)@1 ! WEIGHT = mwght
protein(2)   = g(t6 t1 t6 t2 t6 t5 |ID)@1 ! WEIGHT = pwght
```

6 unknowns for genetic effect

Breeding value for given DIM:

$$S_{dim}^{orig} = \begin{bmatrix} t1_{dim} & 0 & t2_{dim} & 0 & t5_{dim} & 0 \\ 0 & t1_{dim} & 0 & t2_{dim} & 0 & t5_{dim} \end{bmatrix}$$

$$EBV_{dim,i}^{blend} = S_{dim}^{orig} u_i$$

"Solani"-File: Solutions for the Individual Genetic Effects

Column	Description
1	Individual ID
2	Number of Descendants
3	Number of Observations
4	Solution for Trait 1 milk and Factor T1 (ID)
5	Solution for Trait 1 milk and Factor T6 (ID)
6	Solution for Trait 1 milk and Factor T2 (ID)
7	Solution for Trait 1 milk and Factor T6 (ID)
8	Solution for Trait 1 milk and Factor T5 (ID)
9	Solution for Trait 1 milk and Factor T6 (ID)



Blending CLIM file

```

DATAFILE  BLENDFULL.dat
PEDFILE   resfarm.pedi  # same as in original
PARFILE   BLENDING.par  # Updated
TABLEFILE BLEND.cov    # Updated
TABLEINDEX DIM

INTEGER   ID TGRP HERD HYS DIM # TGRP included
REAL      age age2 milk protein mwght pwght # Weights included
MISSING   -99999.0

TRAITGROUP TGRP # TRAITGROUP included

DEFINE D5 - - - - -

MODEL
milk(1)   = CalvAge(age age2) Lac(t1:5) HYS pe(t1 t2 t5|ID) g(t1 t6 t2 t6 t5 t6 |ID)@1
protein(1) = CalvAge(age age2) Lac(t1:5) HYS pe(t1 t2 t5|ID) g(t6 t1 t6 t2 t6 t5 |ID)@1
milk(2)   = CalvAge( - - ) Lac( D5 ) HYS pe(- - - ) g(t1 t6 t2 t6 t5 t6 |ID)@1 ! WEIGHT = mwght
protein(2) = CalvAge( - - ) Lac( D5 ) HYS pe(- - - ) g(t6 t1 t6 t2 t6 t5 |ID)@1 ! WEIGHT = pwght

WITHINBLOCKORDER g pe HYS
RANDOM            pe g
PEDIGREE         g am+p 0.3333

```

Intercept for DRPs modelled in HYS effect

Blending data file

- Trait group column:
 - 1 for test-day records,
 - 2 for drps
- HYS for DRPs always 1.
- DIM for DRPs always 366
- Weights for TD records (tgrp = 1) allways 1

#	ID	TGRP	HERD	HYS	DIM	age	age2	milk	protein	mwght	pwght		
1320041726	1	2	20202403	298	0.334	0.11	27.97	1.23	1	1	#	TDREC	
1320041726	1	2	20202403	305	0.334	0.11	27.17	1.20	1	1	#	TDREC	
1210425818	2	99		1	366	0	0	2078.24	39.96	1.48	1.39	#	DRPS
1210425820	2	99		1	366	0	0	-1824.14	-69.07	1.46	1.37	#	DRPS



Blending parfile

- Variance components for PE and GE the same as in original
- Residual variance

Original residual covariance for test-day observations

```
Residual Co-Variance Matrix
-----
1      2.48999
2      0.934845E-01  0.407896E-02
3      0.00000      0.00000      522060.
4      0.00000      0.00000      18225.0      672.782
```

Residual covariance matrix is for 305d observation and includes variation due to PE + residual

$$\mathbf{R}_{305} = \mathbf{S}\mathbf{K}_p\mathbf{S}' + 305 \times \mathbf{K}_e$$

$$\mathbf{S} = \begin{bmatrix} 215.7 & 0 & -58.1 & 0 & 16.0 & 0 \\ 0 & 215.7 & 0 & -58.1 & 0 & 16.0 \end{bmatrix}$$

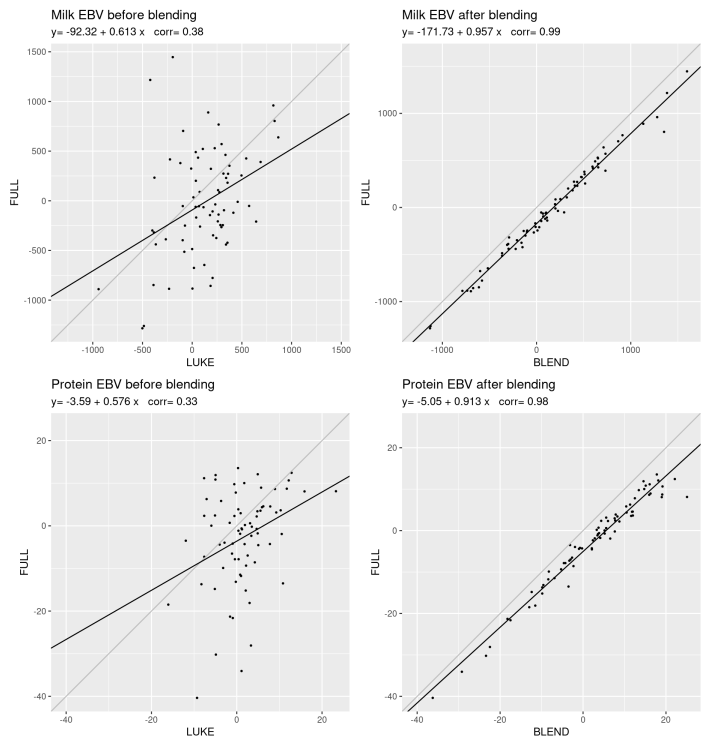
Blending covariable file

DIM	t1	t2	t3	t4	t5	t6
363	0.707106781186548	1.21113659504279	1.52872700590177	1.74783261354842	1.31085650470662e-08	0
364	0.707106781186548	1.21794073321719	1.55485971712122	1.80889999717125	1.2469252785751e-08	0
365	0.707106781186548	1.22474487139159	1.58113883008419	1.87082869338697	1.18611201513438e-08	0
366	215.667568261897	-58.1073400093565	-50.546665969895	-31.5044729042412	15.9686571167008	0

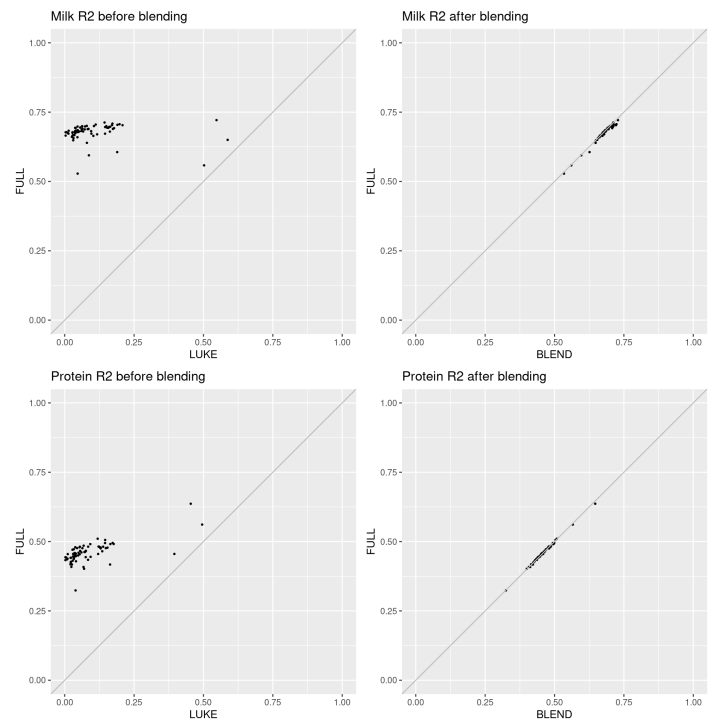
- Covariables 1-5 for DRP are calculated as: $t_{i;366} = \sum_{dim=8}^{312} t_{i,dim}$

Results for blended animals

EBVS



Reliabilities



Mu estimates:
-169.23 -4.8080



Conclusion, example

- The required steps were illustrated
- Blending method works as expected
- When blending MACE information, R2 for MACE evaluation is usually much higher compared to this example
- R2 for domestic evaluation can be low
- 3 animals had negative
$$Weight = ERC_{ext} - ERC_{dom}$$
- Those were omitted from blening
- In the example, 305d DRP for traits were

Alternative 2

Alternative 2: Use covariance functions

```

MODEL
milk(1)      = CalvAge(age age2) Lac(t1:5) HYS pe(t1 t2 t5|ID) g(t7:12|ID)@1
protein(1)   = CalvAge(age age2) Lac(t1:5) HYS pe(t1 t2 t5|ID) g(t13:18|ID)@1
#DRPS
milk(2)      = CalvAge( - - ) Lac( D5 ) HYS pe(- - - ) g(t7:12 |ID)@1 ! WEIGHT = mwght
protein(2)   = CalvAge( - - ) Lac( D5 ) HYS pe(- - - ) g(t13:18 |ID)@1 ! WEIGHT = pwght

```

- Here $\text{var}(a) = I$
- The covariance structure between random effects is explained by covariables
- Covariance functions are used in Nordic test-day model
- Covariables for DRPs: $t_{i;366} = \sum_{dim=8}^{312} t_{i,dim}, i = 7, \dots, 18$
- First steps are the same as for alternative 1



How to get 305d variance components

How to get variance components for EBVs or indices from apax99

```

• # AccurType: Reliability method
• S # Will calculate VC for indices defined
• # MaxNonZ: Number of non-zeros in sparse matrix
• 100000
• # firstDIM: DIM of first row in covariable table
• 5
• # PHImatrix: For each traits: first DIM, number of DIMs, step
size
• 5 305 1
• 5 305 1
• # OriginalDir: MiX99 directives file
• -
• # NumBVs: Number of breeding values for which r2 should be
approximated
• 2
• # Weights: Weights applied for the breeding values
• 1 0
• 0 1
• # H2calc: Random effects used for heritability calculation
• 1 # also 1st random effect (pe) is included
• # JFilter: Within-block effect number from which approximate
absorption is used
• # 1 2 3
• # g pe HYS
• 4

```

- Specify method "S"
- It will calculate genetic and residual variance based on PHImatrix and Weights.
- In beta testing at the moment!

```

>> #-----
>> # AccurType: Reliability method
>> S # Tier and Meyer

Produce variances of random effects to a file,
and G0 and R0 variance matrices to their own files.

```

```

[lst102@sorvi-nd3 tmp]$ cat APAX_G0sum_in_H2.txt
475409.7078291422 9933.058504380109
9933.058504380109 345.9960780403417
[lst102@sorvi-nd3 tmp]$ cat APAX_R0sum_in_H2.txt
522059.5966581812 18225.04283763134
18225.04283763134 672.7818480446911

```



Conclusions

Conclusions

- Alternative 1 and 2 have been successfully implemented for multitrait test-day evaluations
- DRPs can be blended also to single step models
- In that case, remember that ERC and DRP calculation assumes simple animal model!
- If domestic ERC or DRP go to zero for an animal, it is best to set:

$$ERC_{dom} = 0,$$

$$DRP_{blend} = \frac{ERC_{ext} * DRP_{ext} - 0 * DRP_{dom}}{ERC_{ext} - 0} = DRP_{ext},$$

$$Weight = ERC_{ext}$$



Discussion