

Setting-up and solving a simple (and not so simple) RRM with MiX99

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MiX99 course: test-day models and single step genomic prediction

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Introduction to MiX99 and test-day models

- Simple 1 trait test-day model:
 - CLIM, data structure, parameter files, pedigree, covariable file, residual classes

Multitrait test-day models

- 1 lactation 3 trait test-day model:
 - Ordering of random effects and variance components.
 - Residual covariance between observations.
- Full 3 lactations 3 traits test-day model using trait groups:
 - Trait groups.
 - DATASORT, WITHINBLOCKORDER, sorting data and pedigree



Simple 1 trait test-day model

- Let's consider test-day model for 1 lactation milk animal a , dim d , herd test-day h

Fixed Lactation curve
3rd order Legendre
polynomial

Random
HTD

Random PE
2nd order +
Wilmink

Breeding value
2nd order +
Wilmink

$$milk_{adh} = \sum_{i=1}^4 l_i(d)\beta_i + HTD_h + \sum_{i=1}^4 c_i(d)pe_{ai} + \sum_{i=1}^4 c_i(d)u_{ai} + \epsilon_{adh}$$

Residual

Variance components

| HTD | |
|----------|-------|
| | 0.413 |
| Residual | |
| | 1.104 |

| PE | | | | |
|----|--------|-------|--------|--------|
| | 3.028 | 0.033 | 0.088 | -2.249 |
| | 0.033 | 1.796 | 0.197 | 1.837 |
| | 0.088 | 0.197 | 1.057 | -2.521 |
| | -2.249 | 1.837 | -2.521 | 17.058 |

| GE | | | | |
|----|--------|-------|--------|--------|
| | 2.552 | 0.161 | -0.227 | -0.812 |
| | 0.161 | 0.958 | 0.084 | 0.526 |
| | -0.227 | 0.084 | 0.200 | -0.120 |
| | -0.812 | 0.526 | -0.120 | 2.390 |

CLIM for the model

```

TITLE    MT random regression TDM

DATAFILE lact1.dat
INTEGER  HERD ANI LACT HTD HC2Y C2Y AGE DCC DDRY DIM ResClass
REAL     MILK PROTEIN FAT
MISSING  -64.0

PEDFILE  TD9.ped

PEDIGREE G am

TABLEFILE LG3_W004.cov
TABLEINDEX DIM

PARFILE  milk1.par

TMPDIR   tmpMiX

MODEL

MILK   = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)

RANDOM    HTD PE G
#         1    2    3

```

```

# Title for the model. For the reader of CLIM file.

# DATAFILE location.
# Integer columns must be first, then REAL columns.
# Only first 32 characters of the names are used!!
# Missing value is -64. Data file can contain extra columns.

# Genetic effect is named G. Animal model pedigree without
# unknown parent groups is used.
# UPG can be specified as am+p, and random am+p 0.333

# Values for Legendre polynomials and wilmk for each test-day are
# stored to separate file.
# Integer variable DIM is a key between data and tablefile.

# PARFILE location.

# TMPDIR defines the directory for temporary files.

# Model definition.

# Model has fixed lactation curve and
# 4 random effects: 1=HTD ,2=PE, 3=G, and 4= Residual

```

Data, covariable, and parameter table files

head lact1.dat

```
2669 457 1 128191 266903 3 7 1 5 20 1 15.7575 11.8857 9.8591
2669 457 1 128192 266903 3 7 1 5 48 3 20.0062 15.5429 11.3455
2669 457 1 128193 266903 3 7 1 5 75 5 17.2425 12.6143 9.2864
2669 457 1 128194 266903 3 7 1 5 97 6 16.7475 12.6286 10.4727
2669 457 1 128195 266903 3 7 1 5 123 7 16.1287 12.5286 9.6136
2669 457 1 128196 266903 3 7 1 5 158 7 13.1175 10.2 8.2091
2669 457 1 128197 266903 3 7 1 5 181 8 14.4994 11.9286 9.6955
2669 457 1 128198 266903 3 7 1 5 222 9 15.51 13.1143 9.2455
2669 457 1 128199 266903 3 7 1 5 250 9 14.3756 12.1571 9.4091
2669 457 1 128200 266903 3 7 1 5 291 10 12.4987 10.5714 9.6409
```

head LG3_W004.cov

| #DIM | t1 | t2 | t3 | t4 | t5 |
|------|---------|----------|---------|----------|---------|
| 8 | 1.00000 | -1.22474 | 1.58114 | -1.87083 | 0.72615 |
| 9 | 1.00000 | -1.21788 | 1.55464 | -1.80838 | 0.69768 |
| 10 | 1.00000 | -1.21102 | 1.52829 | -1.74681 | 0.67032 |
| 11 | 1.00000 | -1.20416 | 1.50209 | -1.68611 | 0.64404 |
| 12 | 1.00000 | -1.19730 | 1.47603 | -1.62628 | 0.61878 |
| 13 | 1.00000 | -1.19044 | 1.45013 | -1.56731 | 0.59452 |
| 14 | 1.00000 | -1.18358 | 1.42438 | -1.50919 | 0.57121 |
| 15 | 1.00000 | -1.17672 | 1.39877 | -1.45193 | 0.54881 |
| 16 | 1.00000 | -1.16985 | 1.37331 | -1.39551 | 0.52729 |
| 17 | 1.00000 | -1.16299 | 1.34800 | -1.33993 | 0.50662 |

Parfile format is:

Effect number, i, j, val.

Only lower or upper part of the covariance matrix is needed.

First column in covariable table is the key column.

cat milk1.par

```
# HTD
1 1 1 0.41327678
# PE
2 1 1 3.0284761
2 1 2 0.032571576
2 2 2 1.7962894
2 1 3 0.088109073
2 2 3 0.19685252
2 3 3 1.0572924
2 1 4 -2.2488402
2 2 4 1.8372373
2 3 4 -2.5214496
2 4 4 17.057958
# GEN
3 1 1 2.5521561
3 1 2 0.16126933
3 2 2 0.95818471
3 1 3 -0.22728278
3 2 3 0.083704976
3 3 3 0.20033431
3 1 4 -0.81232366
3 2 4 0.5259268
3 3 4 -0.11959052
3 4 4 2.3896462
#RESID
4 1 1 1.1035821
```

Running preprocessor

- Run preprocessor

```
$ mix99i milk1.clm
```

- Mix99i prints model information to `Mix99.lst` file and standard output.
- After successful execution, `OK_mix99i` file is created.
- It can be used in pipelines to check for errors.
- It contains timing information.
- **ALWAYS check** at least DATA DESCRIPTION from `Mix99.lst`!!
- Also check the standard output!

```
DATA DESCRIPTION
=====
PEDIGREE:
Number of Pedigree Records Processed ..... 30082
Number of Blocks in Pedigree File ..... 4
Number of Blocks in Pedigree and Data File ..... 22

DEPENDENT VARIABLES:
TR TR-NAME      N-OBS      MEAN      SD      MINIMUM      MAXIMUM
-----
 1  MILK          178104    15.926    3.6973    0.18560     33.495

Number of Animals..... 30082
Number of UPG..... 0
Number of Animals with data records..... 19928
Number of Data Records Processed ..... 178104
Total Number of Observations ..... 178104
Number of Trait Patterns ..... 1
Maximum Number of Trait Patterns ..... 1

COVARIABLES READ FROM COVARIABLE TABLE:
tCOVARIABLE      N      MEAN      SD      MINIMUM      MAXIMUM
-----
 1          358      1.00000    0.00000    1.00000     1.00000
 2          358      0.00000    0.710077   -1.22474     1.22474
 3          358      0.442905E-02  0.712055   -0.790550     1.58114
 4          358      0.00000    0.714082   -1.87083     1.87083
 5          358      0.517294E-01  0.128560    0.00000     0.726150

Number of residual (co)variance matrices ... 1
```

Running solver

- Run solver with default settings and produce default solution files

```
$ mix99s -s
```

Or give own settings:

```
$ mix99s < solver.slv
```

Example of MiX99 solver option file:

```
# RAM: RAM demand: X=large (mix99p only), H=high, M=medium, L=low; nt 10 CPU threads
H   nt 10
# STOP: Max. num. iterations, Stopping criterion, Convergence indicator, enforce
      5000          5.0e-5          d          f
# RESID: Calculate residuals? (Y/N)
      N
# VALID: N=none, P=prediction, S=sum of effects, Y=YD, D=DYD, I=IDD
      N
# VAROPT: adjust for heterogeneous variance / variance components: (N, E, S, C)
      N
# SOLTYP: type of solution files? (Y,N,A,H)
      Y
```

- After successful execution, OK_mix99s file is created
- It contains timing and other technical information.
- The other information is written to the standard output.
- Iteration statistics are also written to Conlog

Output of the solver

- Solver output has the following parts
 - version information,
 - preconditioner information,
 - solver parameters,
 - iteration statistics,
 - solutions,
 - description of solution files.

Convergence

PCG method solves a system of equations: $\mathbf{M}^{-1}\mathbf{C}\mathbf{s} = \mathbf{M}^{-1}\mathbf{r}$,

The first four [convergence indicators](#) are norms that can be selected as the [stopping criterion](#) of the iteration. For describing these norms, we define that C represents the coefficient matrix of [MME](#), $\hat{\mathbf{s}}^{(k)}$ the vector of solutions at round k , \mathbf{r} the right hand side of the [MME](#), and M^{-1} the inverse of the [preconditioner](#) matrix, which approximates the inverse of C . The four norms are:

$$ca_{(k)} = \sqrt{\frac{(\mathbf{r} - C\hat{\mathbf{a}}^{(k)})^T (\mathbf{r} - C\hat{\mathbf{a}}^{(k)})}{(\mathbf{r}_a)^T (\mathbf{r}_a)}},$$

$$cr_{(k)} = \sqrt{\frac{(\mathbf{r} - C\hat{\mathbf{s}}^{(k)})^T (\mathbf{r} - C\hat{\mathbf{s}}^{(k)})}{(\mathbf{r})^T (\mathbf{r})}},$$

$$cm_{(k)} = \sqrt{\frac{(\mathbf{r} - C\hat{\mathbf{s}}^{(k)})^T M^{-1} (\mathbf{r} - C\hat{\mathbf{s}}^{(k)})}{(\mathbf{r})^T M^{-1} (\mathbf{r})}},$$

$$cd_{(k)} = \sqrt{\frac{(\hat{\mathbf{s}}^{(k)} - \hat{\mathbf{s}}^{(k-1)})^T (\hat{\mathbf{s}}^{(k)} - \hat{\mathbf{s}}^{(k-1)})}{(\hat{\mathbf{s}}^{(k)})^T (\hat{\mathbf{s}}^{(k)})}},$$

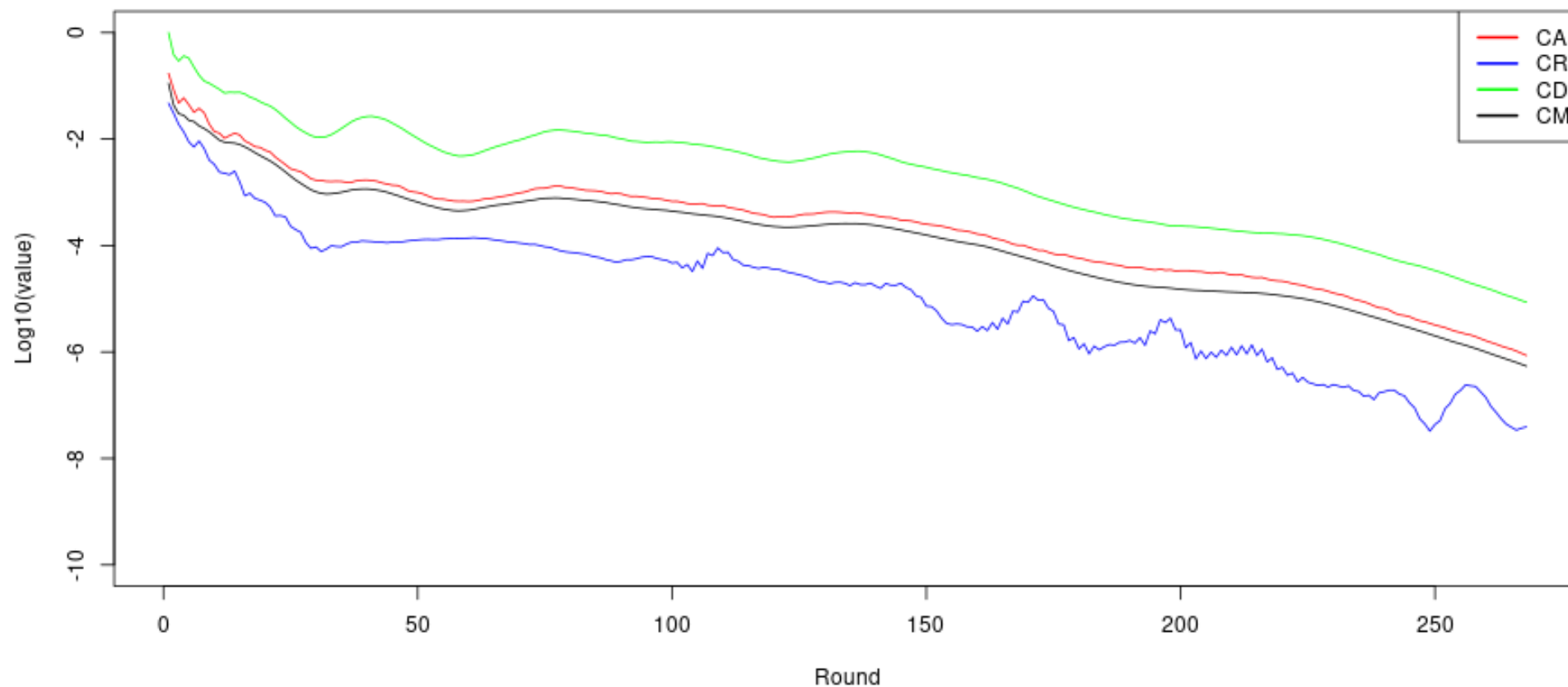
where $ca_{(k)}$ is the relative difference between left-hand side and right-hand side (residual) of the part of the [MME](#) which includes the equations of the additive genetic animal effects; $cr_{(k)}$ is the relative residual of all effects of the [MME](#); $cm_{(k)}$ is the [preconditioned](#) relative residual of the [MME](#); and $cd_{(k)}$ is the relative difference between solutions from consecutive iterations.

```

Iteration Statistics
-----
Convergence Indicators
-----
ROUND      CA          CR          CM          CD          MAX.CHA.
-----
Solution vector will be initialized to be zero
rhs' * rhs = 7055565005107.81
animal rhs' * rhs = 403145030.811337
-----
0      1.000      1.000      1.000      0.000      0.000
1      0.1711     0.4712E-01  0.1113     1.000     11.92
2      0.8467E-01 0.3065E-01  0.4364E-01 0.3838    -7.684
3      0.4730E-01 0.1884E-01  0.2994E-01 0.2923    -5.683
4      0.5941E-01 0.1363E-01  0.2785E-01 0.3655    -8.371
5      0.4402E-01 0.8819E-02  0.2247E-01 0.3300    -8.181
6      0.3186E-01 0.7084E-02  0.2134E-01 0.2249    -7.121
7      0.3779E-01 0.9208E-02  0.1768E-01 0.1583     5.188

```

Convergence indicators



Solution files: Solreg(Lactcurve) and Solr01 (HTD)

```
"Solreg"-File: Solutions for General Regressions

Column | Description
-----|-----
 1      | Trait Number
 2      | Regression Number within Trait
 3      | Solution
 4      | Name of Trait
 5      | Name of Covariable

head Solreg
  Trt Reg-No Solution   Trait Covariable
  --- --
  1   1  12.914     MILK   T1
  1   2  -3.2704    MILK   T2
  1   3  -1.4560    MILK   T3
  1   4  -0.37894   MILK   T4
```

```
"Solr01"-File: Solutions for Random Effect           1   HTD

Column | Description
-----|-----
 1      | Level Code
 2      | Number of Observations
 3      | Solution for Trait 1 MILK           and Factor HTD

head Solr01
128191      24  0.28677E-01
128192      24  0.39835
128193      23  1.2022
128194      23  1.2443
128195      22  0.33293
128196      19 -1.1561
128197      18 -0.23096
128198      24  0.72250
128199      26 -0.24111E-02
128200      29 -0.86877
```



Solution files Solr02 (PE) and Solani (EBV)

"Solr02"-File: Solutions for Random Effect 2 ANI

Column | Description

```
-----
 1      Level Code
 2      Number of Observations
 3      Solution for Trait 1 MILK      and Factor ANI
 3      Solution for Trait 1 MILK      and Factor T2(ANI)
 3      Solution for Trait 1 MILK      and Factor T3(ANI)
 3      Solution for Trait 1 MILK      and Factor T5(ANI)
```

head Solr02

```
 74      9  0.94771      -0.99717      -0.69752      -1.2687
 76     10  0.97040       0.54684      -0.56101       3.1141
114     10 -1.3708      -0.65507E-01  0.70238      -4.1157
143      9 -0.30097       0.72059      -0.17489E-01  1.0715
145      9  0.43337E-01 -0.75115      -0.58631      -1.7284
146      9  0.51494       0.27084       0.81475      -0.11436
157     11 -0.76751       0.26545       1.0902      -2.5653
163     11 -3.5028       0.99895      -0.51525       0.14865
165     12  2.1606       3.4978       0.74588       7.5912
226      1 -0.46705       0.17018       0.80096E-01 -1.0483
```

"Solani"-File: Solutions for Genetic Animal Effect

Column | Description

```
-----
 1      Animal ID
 2      Number of Descendants
 3      Number of Observations
 4      Solution for Trait 1 MILK      and Factor ANI
 5      Solution for Trait 1 MILK      and Factor T2(ANI)
 6      Solution for Trait 1 MILK      and Factor T3(ANI)
 7      Solution for Trait 1 MILK      and Factor T5(ANI)
```

head Solani

```
 1      1      0  2.7142      -1.0601       0.30033      -6.3599
 4      1      0  1.5124      -1.5898       0.28041      -6.6436
 7      1      0  2.1711      -0.98947      1.1029      -9.3518
10     1      0  2.3678      -1.0570       0.64300      -6.4742
13     1      0  1.9901      -0.59906      0.33120      -6.9854
16     1      0  1.4587      -0.96090      0.30514      -5.4235
 3      2      0  2.5619      -0.97868      0.83116E-01 -4.9193
20     1      0  2.4667      -0.89210      0.87934      -8.5960
23     1      0  3.0872      -1.3788       0.41423      -6.4622
25     1      0  0.88145     -0.59714      0.29842      -4.8686
```



Add residual variance classes to model

```
TITLE MT random regression TDM
DATAFILE lact1.dat
INTEGER HERD ANI LACT HTD HC2Y C2Y AGE DCC DDRY DIM ResClass
REAL MILK PROTEIN FAT
MISSING -64.0
```

```
PEDFILE TD9.ped
PEDIGREE G am
```

```
TABLEFILE LG3_W004.cov
TABLEINDEX DIM
```

```
RESIDUAL ResClass
RESIDFILE milk1.res
```

```
PARFILE milk1.par
```

```
RANDOM HTD PE G
```

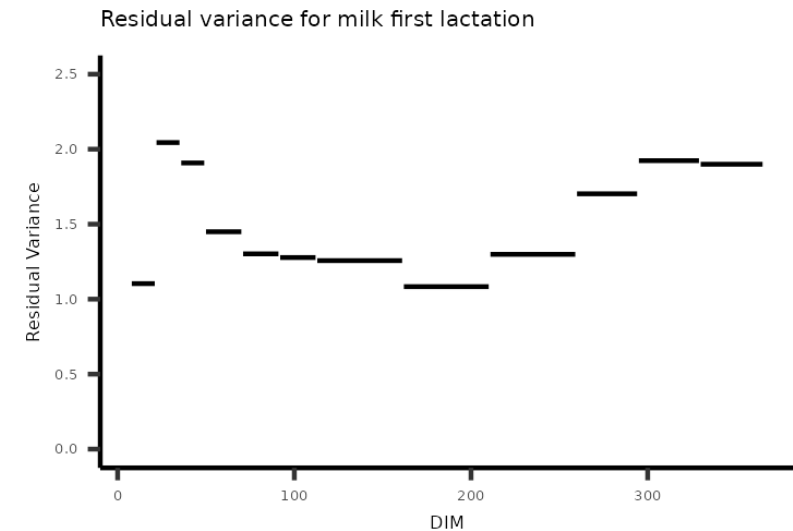
```
TMPDIR tmpMiX
```

```
MODEL
```

```
MILK = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)
```

- Integer column **ResClass** specifies which residual class observation belongs to
- Residual variances for classes are given in own parameter file (milk1.res) with columns:
class number, i,j, value

```
cat milk1.res
1 1 1 1.1035821
2 1 1 2.044039
3 1 1 1.9081354
4 1 1 1.4494774
5 1 1 1.3019862
6 1 1 1.2774864
7 1 1 1.2569414
8 1 1 1.0833885
9 1 1 1.2988478
10 1 1 1.7024491
11 1 1 1.9234934
12 1 1 1.8995175
```



Multiple traits random regression

```

TITLE  MT random regression TDM
DATAFILE lact1.dat
INTEGER  HERD ANI LACT HTD HC2Y C2Y AGE DCC DDRY DIM ResClass
REAL    MILK PROTEIN FAT
MISSING -64.0

PEDFILE  TD9.ped
PEDIGREE G am

TABLEFILE LG3_W004.cov
TABLEINDEX DIM

RESIDUAL ResClass
RESIDFILE lactation1.res

PARFILE  lactation1.par

RANDOM   HTD PE G

TMPDIR  tmpMiX

MODEL

MILK    =  LACcurve (t1 t2 t3 t4) HTD PE (1 t2 t3 t5|ANI)  G(1 t2 t3 t5|ANI)
PROTEIN =  LACcurve (t1 t2 t3 t4) HTD PE (1 t2 t3 t5|ANI)  G(1 t2 t3 t5|ANI)
FAT     =  LACcurve (t1 t2 t3 t4) HTD PE (1 t2 t3 t5|ANI)  G(1 t2 t3 t5|ANI)

```

- Model rows for protein and fat are just copy paste from milk.
- The model effects for each trait are the same.
- Sizes of variance component matrices are
HTD = 3x3, PE=12x12, G=12x12.
- The tricky part is the structure of covariance matrices for PE and G.

Covariance matrices for 1 first lactation model: HTD

- HTD effect has only three random variables for each herd test-day:

$$h_{htd} = \begin{bmatrix} h_{htd; \text{milk}} \\ h_{htd; \text{prot}} \\ h_{htd; \text{fat}} \end{bmatrix},$$

$$\text{var}(h_{htd}) = \begin{bmatrix} 0.41 & 0.31 & 0.20 \\ 0.31 & 0.30 & 0.18 \\ 0.20 & 0.18 & 0.26 \end{bmatrix}$$

```

MILK      = LACcurve (t1 t2 t3 t4) HTD PE (1 t2 t3 t5 |ANI)  G (1 t2 t3 t5 |ANI)
PROTEIN   = LACcurve (t1 t2 t3 t4) HTD PE (1 t2 t3 t5 |ANI)  G (1 t2 t3 t5 |ANI)
FAT       = LACcurve (t1 t2 t3 t4) HTD PE (1 t2 t3 t5 |ANI)  G (1 t2 t3 t5 |ANI)
  
```

Covariance matrices for 1 first lactation model: PE and G

- PE and G have 4 random regression coefficients for each trait.
- MiX99 numbers random variables first by row, then by column.
- Variable numbers are in subscript.
- Ordering of random variables is the same for PE and G.
- Both effects require 78 covariance parameters!

$$pe_i = \begin{bmatrix} p1_i; milk \\ p1_i; prot \\ p1_i; fat \\ pt2_i; milk \\ pt2_i; prot \\ pt2_i; fat \\ pt3_i; milk \\ pt3_i; prot \\ pt3_i; fat \\ pt5_i; milk \\ pt5_i; prot \\ pt5_i; fat \end{bmatrix}$$

| | | | | | |
|----------------|---|------------------------|-----|--|---|
| MILK | = | LACcurve (t1 t2 t3 t4) | HTD | PE (1 ₁ t2 ₄ t3 ₇ t5 ₁₀ ANI) | G (1 ₁ t2 ₄ t3 ₇ t5 ₁₀ ANI) |
| PROTEIN | = | LACcurve (t1 t2 t3 t4) | HTD | PE (1 ₂ t2 ₅ t3 ₈ t5 ₁₁ ANI) | G (1 ₂ t2 ₅ t3 ₈ t5 ₁₁ ANI) |
| FAT | = | LACcurve (t1 t2 t3 t4) | HTD | PE (1 ₃ t2 ₆ t3 ₉ t5 ₁₂ ANI) | G (1 ₃ t2 ₆ t3 ₉ t5 ₁₂ ANI) |

Covariance matrices for 1 first lactation model: PE and G

- PE covariance matrix

```

3.03 0.03 0.09 -2.25 2.16 0.32 0.04 -0.99 1.67 0.18 0.12 -0.70
0.03 1.80 0.20 1.84 0.07 1.26 0.22 0.94 0.06 1.01 0.13 0.72
0.09 0.20 1.06 -2.52 0.08 0.11 0.77 -1.92 0.07 0.10 0.59 -1.53
-2.25 1.84 -2.52 17.06 -1.63 1.31 -1.77 12.13 -1.26 1.03 -1.50 9.78
2.16 0.07 0.08 -1.63 1.61 0.25 0.05 -0.84 1.22 0.16 0.09 -0.56
0.32 1.26 0.11 1.31 0.25 1.01 0.12 0.96 0.22 0.75 0.09 0.63
0.04 0.22 0.77 -1.77 0.05 0.12 0.61 -1.53 0.04 0.11 0.45 -1.16
-0.99 0.94 -1.92 12.13 -0.84 0.96 -1.53 10.30 -0.60 0.60 -1.16 7.42
1.67 0.06 0.07 -1.26 1.22 0.22 0.04 -0.60 1.05 0.08 0.08 -0.59
0.18 1.01 0.10 1.03 0.16 0.75 0.11 0.60 0.08 0.71 0.03 0.84
0.12 0.13 0.59 -1.50 0.09 0.09 0.45 -1.16 0.08 0.03 0.41 -1.16
-0.70 0.72 -1.53 9.78 -0.56 0.63 -1.16 7.42 -0.59 0.84 -1.16 9.01

```

- G covariance matrix

```

2.55 0.16 -0.23 -0.81 1.62 0.31 -0.17 0.02 1.17 0.17 -0.04 0.00
0.16 0.96 0.08 0.53 0.09 0.56 0.10 0.09 0.16 0.42 0.10 -0.29
-0.23 0.08 0.20 -0.12 -0.14 0.03 0.12 -0.15 -0.04 0.05 0.08 -0.06
-0.81 0.53 -0.12 2.39 -0.53 0.21 -0.02 0.97 -0.37 0.14 -0.01 0.47
1.62 0.09 -0.14 -0.53 1.29 0.18 -0.14 -0.17 0.92 0.11 -0.04 -0.16
0.31 0.56 0.03 0.21 0.18 0.39 0.04 0.12 0.20 0.28 0.05 -0.13
-0.17 0.10 0.12 -0.02 -0.14 0.04 0.09 -0.07 -0.07 0.04 0.05 -0.03
0.02 0.09 -0.15 0.97 -0.17 0.12 -0.07 0.88 -0.13 0.05 -0.05 0.61
1.17 0.16 -0.04 -0.37 0.92 0.20 -0.07 -0.13 0.95 0.09 -0.01 -0.40
0.17 0.42 0.05 0.14 0.11 0.28 0.04 0.05 0.09 0.24 0.04 -0.05
-0.04 0.10 0.08 -0.01 -0.04 0.05 0.05 -0.05 -0.01 0.04 0.05 -0.05
0.00 -0.29 -0.06 0.47 -0.16 -0.13 -0.03 0.61 -0.40 -0.05 -0.05 1.47

```

```

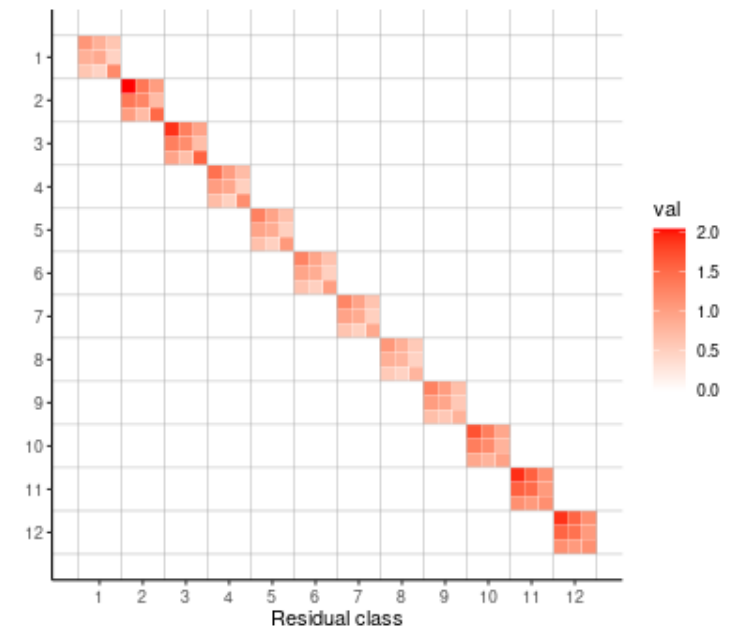
MILK      = LACcurve(t1 t2 t3 t4) HTD PE(11 t24 t37 t510 |ANI) G(11 t24 t37 t510 |ANI)
PROTEIN   = LACcurve(t1 t2 t3 t4) HTD PE(12 t25 t38 t511 |ANI) G(12 t25 t38 t511 |ANI)
FAT       = LACcurve(t1 t2 t3 t4) HTD PE(13 t26 t39 t512 |ANI) G(13 t26 t39 t512 |ANI)

```

Residual covariance in multitrait models

- Residual covariance between traits is applied to observations in the same record (dataline).
- Residual covariance between observations in different data lines is assumed to be 0.
- In this model 12 residual classes are used
- Each residual class has 3x3 covariance matrix.
- In total, $12 * 6 = 72$ covariance parameters for residual
- If the dependency of residuals for different DIMs is needed, it can be modelled using PE like random animal effect.
- However, it is challenging and may cause convergence issues.

Residual covariance matrices for residual classes



Solution files for first lactation model

"Solreg"-File: Solutions for General Regressions

| Column | Description |
|--------|--------------------------------|
| 1 | Trait Number |
| 2 | Regression Number within Trait |
| 3 | Solution |
| 4 | Name of Trait |
| 5 | Name of Covariable |

head Solreg

| Trt | Reg-No | Solution | Trait | Covariable |
|-----|--------|----------|---------|------------|
| 1 | 1 | 12.202 | MILK | T1 |
| 1 | 2 | -2.9059 | MILK | T2 |
| 1 | 3 | -1.1471 | MILK | T3 |
| 1 | 4 | -0.31725 | MILK | T4 |
| 2 | 1 | 10.083 | PROTEIN | T1 |
| 2 | 2 | -1.5859 | PROTEIN | T2 |
| 2 | 3 | -0.87709 | PROTEIN | T3 |
| 2 | 4 | -0.11994 | PROTEIN | T4 |
| 3 | 1 | 8.1223 | FAT | T1 |
| 3 | 2 | -1.3749 | FAT | T2 |
| 3 | 3 | -0.57326 | FAT | T3 |
| 3 | 4 | -0.22634 | FAT | T4 |

"Solr01"-File: Solutions for Random Effect

1 HTD

| Column | Description |
|--------|---|
| 1 | Level Code |
| 2 | Number of Observations |
| 3 | Solution for Trait 1 MILK and Factor HTD |
| 4 | Solution for Trait 2 PROTEIN and Factor HTD |
| 5 | Solution for Trait 3 FAT and Factor HTD |

head Solr01

| | | | | |
|--------|----|--------------|--------------|-------------|
| 128191 | 24 | -0.17965E-01 | -0.52881E-01 | -0.30661 |
| 128192 | 24 | 0.41037 | 0.36624 | 0.46464E-01 |
| 128193 | 23 | 1.1574 | 0.76298 | 0.27929 |
| 128194 | 23 | 1.2631 | 1.2177 | 0.89137 |
| 128195 | 22 | 0.33380 | 0.44346 | -0.27413 |
| 128196 | 19 | -1.0824 | -0.95181 | -0.95102 |
| 128197 | 18 | -0.15666 | -0.18544 | -0.20665 |
| 128198 | 24 | 0.62112 | 0.61441 | 0.77612E-01 |
| 128199 | 26 | -0.20994E-01 | 0.18393 | 0.22617 |
| 128200 | 29 | -0.89371 | -0.46519 | 0.25696 |

Solution files for first lactation model

"Solr02"-File: Solutions for Random Effect

2 ANI

Column | Description

| Column | Description |
|--------|---|
| 1 | Level Code |
| 2 | Number of Observations |
| 3 | Solution for Trait 1 MILK and Factor ANI |
| 4 | Solution for Trait 2 PROTEIN and Factor ANI |
| 5 | Solution for Trait 3 FAT and Factor ANI |
| 3 | Solution for Trait 1 MILK and Factor T2(ANI) |
| 4 | Solution for Trait 2 PROTEIN and Factor T2(ANI) |
| 5 | Solution for Trait 3 FAT and Factor T2(ANI) |
| 3 | Solution for Trait 1 MILK and Factor T3(ANI) |
| 4 | Solution for Trait 2 PROTEIN and Factor T3(ANI) |
| 5 | Solution for Trait 3 FAT and Factor T3(ANI) |
| 3 | Solution for Trait 1 MILK and Factor T5(ANI) |
| 4 | Solution for Trait 2 PROTEIN and Factor T5(ANI) |
| 5 | Solution for Trait 3 FAT and Factor T5(ANI) |

head Solr02

```

74 9 0.99156 0.56274 0.45377 -0.80898 -0.26952 -0.31023 ...
76 10 1.1189 0.99433 0.74311 0.45577 0.63977 0.21155 ...
114 10 -1.3312 -0.91677 -0.61445 0.49116E-01 -0.62644E-01 -0.78481E-02 ...
143 9 -0.35951 -0.19074 -0.23220 0.65900 0.10653 0.35289 ...
145 9 0.25916 0.21268 0.31821 -0.57542 -0.25931 -0.20978 ...
146 9 0.54251 0.62274 0.38799 0.31331 0.13336 0.28073 ...
157 11 -0.92067 -0.79038 -0.44410 -0.73007E-01 -0.31008 -0.29550 ...
163 11 -3.6774 -2.6107 -2.0315 0.86495 0.23697 0.12217 ...
165 12 1.8915 1.0388 1.0100 3.1267 2.4545 1.6869 ...
226 1 -0.55895 -0.37087 -0.32768 0.20016 0.51922E-01 0.53863E-01 ...

```

"Solani"-File: Solutions for Genetic Animal Effect

Column | Description

| Column | Description |
|--------|---|
| 1 | Animal ID |
| 2 | Number of Descendants |
| 3 | Number of Observations |
| 4 | Solution for Trait 1 MILK and Factor ANI |
| 5 | Solution for Trait 2 PROTEIN and Factor ANI |
| 6 | Solution for Trait 3 FAT and Factor ANI |
| 7 | Solution for Trait 1 MILK and Factor T2(ANI) |
| 8 | Solution for Trait 2 PROTEIN and Factor T2(ANI) |
| 9 | Solution for Trait 3 FAT and Factor T2(ANI) |
| 10 | Solution for Trait 1 MILK and Factor T3(ANI) |
| 11 | Solution for Trait 2 PROTEIN and Factor T3(ANI) |
| 12 | Solution for Trait 3 FAT and Factor T3(ANI) |
| 13 | Solution for Trait 1 MILK and Factor T5(ANI) |
| 14 | Solution for Trait 2 PROTEIN and Factor T5(ANI) |
| 15 | Solution for Trait 3 FAT and Factor T5(ANI) |

head Solani

```

1 1 0 3.5729 2.4194 1.5256 -1.1981 -0.27947 -0.26116 ...
4 1 0 2.3634 1.2145 1.4067 -1.6518 -0.49062 -0.52810 ...
7 1 0 2.7368 1.9404 2.0480 -1.2383 -0.25751 -0.18850 ...
10 1 0 2.8937 1.8512 1.3084 -1.0411 -0.19864 -0.12744 ...
13 1 0 2.7386 1.6019 1.0450 -0.70816 -0.47939E-01 -0.77933E...
16 1 0 1.9904 0.79604 0.58778 -1.0534 -0.32418 -0.35039 ...
3 2 0 3.2071 2.1734 1.2498 -1.1350 -0.15555 -0.16690 ...
20 1 0 3.3834 2.0699 1.3694 -1.1374 -0.23903 -0.31426 ...
23 1 0 3.8460 3.0194 1.2792 -1.6866 -0.45594 -0.12098 ...
25 1 0 1.2965 0.79091 0.91466 -0.82220 -0.18907 -0.18593 ...

```

Trait groups

- The previous models had only first lactation traits
- Test-day models can have more than 1 lactation.
- Usually milk, protein and fat yields for each lactation are considered as own traits
- Trait group feature in MiX99 uses one integer column in the data to specify which trait group the record is from.
- In the model line, trait group number is within parentheses after trait number.
- In general, effects and traits can vary between trait groups.

Test-day model for 3 lactations

- Integer column **LACT** specifies the lactation
- In this model, each trait has the same effect structure.
- No need to define lactation specific columns for observations!

```

INTEGER   HERD ANI LACT HTD HC2Y C2Y AGE DCC DDRY DIM ResClass
REAL      MILK PROTEIN FAT

RANDOM     HTD PE G
TRAITGROUP LACT

MODEL
MILK (1)   = LACcurve (t1 t2 t3 t4) HTD PE (1 t2 t3 t5|ANI) G (1 t2 t3 t5|ANI)
PROTEIN (1) = LACcurve (t1 t2 t3 t4) HTD PE (1 t2 t3 t5|ANI) G (1 t2 t3 t5|ANI)
FAT (1)     = LACcurve (t1 t2 t3 t4) HTD PE (1 t2 t3 t5|ANI) G (1 t2 t3 t5|ANI)
MILK (2)   = LACcurve (t1 t2 t3 t4) HTD PE (1 t2 t3 t5|ANI) G (1 t2 t3 t5|ANI)
PROTEIN (2) = LACcurve (t1 t2 t3 t4) HTD PE (1 t2 t3 t5|ANI) G (1 t2 t3 t5|ANI)
FAT (2)     = LACcurve (t1 t2 t3 t4) HTD PE (1 t2 t3 t5|ANI) G (1 t2 t3 t5|ANI)
MILK (3)   = LACcurve (t1 t2 t3 t4) HTD PE (1 t2 t3 t5|ANI) G (1 t2 t3 t5|ANI)
PROTEIN (3) = LACcurve (t1 t2 t3 t4) HTD PE (1 t2 t3 t5|ANI) G (1 t2 t3 t5|ANI)
FAT (3)     = LACcurve (t1 t2 t3 t4) HTD PE (1 t2 t3 t5|ANI) G (1 t2 t3 t5|ANI)

head lact3.dat
2669 457 1 128191 266903 3 7 1 5 20 1 15.7575 11.8857 9.8591
2669 457 1 128192 266903 3 7 1 5 48 3 20.0062 15.5429 11.3455
2669 457 1 128193 266903 3 7 1 5 75 5 17.2425 12.6143 9.2864
2669 457 1 128194 266903 3 7 1 5 97 6 16.7475 12.6286 10.4727
2669 457 1 128195 266903 3 7 1 5 123 7 16.1287 12.5286 9.6136
2669 457 1 128196 266903 3 7 1 5 158 7 13.1175 10.2 8.2091
2669 457 1 128197 266903 3 7 1 5 181 8 14.4994 11.9286 9.6955

```

Test-day model for 3 lactations: variance components

- Trait groups don't have direct effect to covariance matrices.
- The order of the model lines defines the order of variance components.
- The sizes of covariance matrices:
 - HTD = 9x9
 - PE = 36x36
 - G = 36x36
 - Residual has 12 classes. Each class has 9x9 covariance matrix. Covariance between lactations 0.
- In total 1593 variance components!

```

MILK (1)   = .. HTD1 PE (11 t210 t319 t528 |ANI) G (11 t210 t319 t528 |ANI)
PROTEIN (1) = .. HTD2 PE (12 t211 t320 t529 |ANI) G (12 t211 t320 t529 |ANI)
FAT (1)     = .. HTD3 PE (13 t212 t321 t530 |ANI) G (13 t212 t321 t530 |ANI)
MILK (2)   = .. HTD4 PE (14 t213 t322 t531 |ANI) G (14 t213 t322 t531 |ANI)
PROTEIN (2) = .. HTD5 PE (15 t214 t323 t532 |ANI) G (15 t214 t323 t532 |ANI)
FAT (2)     = .. HTD6 PE (16 t215 t324 t533 |ANI) G (16 t215 t324 t533 |ANI)
MILK (3)   = .. HTD7 PE (17 t216 t325 t534 |ANI) G (17 t216 t325 t534 |ANI)
PROTEIN (3) = .. HTD8 PE (18 t217 t326 t535 |ANI) G (18 t217 t326 t535 |ANI)
FAT1 (3)   = .. HTD9 PE (19 t218 t327 t536 |ANI) G (19 t218 t327 t536 |ANI)

```

Genetic solutions for 9 trait model

```
"Solani"-File: Solutions for Genetic Animal Effect
```

| Column | Description |
|--------|---|
| 1 | Animal ID |
| 2 | Number of Descendants |
| 3 | Number of Observations |
| 4 | Solution for Trait 1 MILK and Factor ANI |
| 5 | Solution for Trait 2 PROTEIN and Factor ANI |
| 6 | Solution for Trait 3 FAT and Factor ANI |
| 7 | Solution for Trait 4 MILK and Factor ANI |
| 8 | Solution for Trait 5 PROTEIN and Factor ANI |
| 9 | Solution for Trait 6 FAT and Factor ANI |
| 10 | Solution for Trait 7 MILK and Factor ANI |
| 11 | Solution for Trait 8 PROTEIN and Factor ANI |
| 12 | Solution for Trait 9 FAT and Factor ANI |
| 13 | Solution for Trait 1 MILK and Factor T2(ANI) |
| 14 | Solution for Trait 2 PROTEIN and Factor T2(ANI) |
| 15 | Solution for Trait 3 FAT and Factor T2(ANI) |
| 16 | Solution for Trait 4 MILK and Factor T2(ANI) |
| 17 | Solution for Trait 5 PROTEIN and Factor T2(ANI) |
| 18 | Solution for Trait 6 FAT and Factor T2(ANI) |
| 19 | Solution for Trait 7 MILK and Factor T2(ANI) |
| 20 | Solution for Trait 8 PROTEIN and Factor T2(ANI) |
| 21 | Solution for Trait 9 FAT and Factor T2(ANI) |
| 22 | Solution for Trait 1 MILK and Factor T3(ANI) |
| 23 | Solution for Trait 2 PROTEIN and Factor T3(ANI) |
| 24 | Solution for Trait 3 FAT and Factor T3(ANI) |
| 25 | Solution for Trait 4 MILK and Factor T3(ANI) |
| 26 | Solution for Trait 5 PROTEIN and Factor T3(ANI) |
| 27 | Solution for Trait 6 FAT and Factor T3(ANI) |
| 28 | Solution for Trait 7 MILK and Factor T3(ANI) |
| 29 | Solution for Trait 8 PROTEIN and Factor T3(ANI) |
| 30 | Solution for Trait 9 FAT and Factor T3(ANI) |
| 31 | Solution for Trait 1 MILK and Factor T5(ANI) |
| 32 | Solution for Trait 2 PROTEIN and Factor T5(ANI) |
| 33 | Solution for Trait 3 FAT and Factor T5(ANI) |
| 34 | Solution for Trait 4 MILK and Factor T5(ANI) |
| 35 | Solution for Trait 5 PROTEIN and Factor T5(ANI) |
| 36 | Solution for Trait 6 FAT and Factor T5(ANI) |
| 37 | Solution for Trait 7 MILK and Factor T5(ANI) |
| 38 | Solution for Trait 8 PROTEIN and Factor T5(ANI) |
| 39 | Solution for Trait 9 FAT and Factor T5(ANI) |

- Solani file has 39 columns
- When using the columns in Solani, care must be taken to ensure that the calculations are correct

DATASORT

```

INTEGER  HERD ANI LACT HTD HC2Y C2Y AGE DCC DDRY DIM ResClass
REAL     MILK PROTEIN FAT

RANDOM    HTD  PE G
TRAITGROUP LACT

MODEL
MILK(1)   = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)
PROTEIN(1) = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)
FAT(1)    = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)
MILK(2)   = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)
PROTEIN(2) = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)
FAT(2)    = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)
MILK(3)   = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)
PROTEIN(3) = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)
FAT(3)    = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)

DATASORT BLOCK=HERD PEDIGREECODE=ANI

WITHINBLOCKORDER G PE HTD

```

- Ordering data AND pedigree by block can speed up solving the model
- Natural block in test-day models are herd – code
- Pedigree must contain block code for each animal (also for pedigree animals) as 4th column.
- Integer variable for block code and pedigree code is given in DATASORT BLOCK command

WITHINBLOCKORDER

```

INTEGER   HERD ANI LACT HTD HC2Y C2Y AGE DCC DDRY DIM ResClass
REAL      MILK PROTEIN FAT

RANDOM     HTD PE G
TRAITGROUP LACT

MODEL
MILK(1)   = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)
PROTEIN(1) = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)
FAT(1)    = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)
MILK(2)   = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)
PROTEIN(2) = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)
FAT(2)    = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)
MILK(3)   = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)
PROTEIN(3) = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)
FAT(3)    = LACcurve(t1 t2 t3 t4) HTD PE(1 t2 t3 t5|ANI) G(1 t2 t3 t5|ANI)

DATASORT  BLOCK=HERD PEDIGREECODE=ANI

WITHINBLOCKORDER G PE HTD

```

- When BLOCK code is given, some of the effects can be classified as “within block” effects using WITHINBLOCKORDER statement.
- Within block effect levels usually have interaction with block variable, like HERD x TEST-DAY, or whose observations mainly belong to one block, like animal effects .
- Genetic effect must be always the first within block effect.
- WITHINBLOCKORDER is needed for R2 calculation.

Sorting data and pedigree for DATASORT

```

INTEGER  HERD ANI LACT HTD HC2Y C2Y AGE DCC DDRY DIM ResClass
REAL     MILK PROTEIN FAT

TRAITGROUP LACT

DATASORT  BLOCK=HERD PEDIGREECODE=ANI

WITHINBLOCKORDER G PE HTD

```

- MiX99 checks if the data is in correct order

DATA

```

#           BLOCK  animal  tgrp  DIM
>sort -n -k1,1 -k2,2 -k3,3 -k10,10 lact3.dat > lact3.s

```

Pedigree

```

#           BLOCK  ANIMAL
>sort -n -k4,4 -k1,1 TD9.ped

```

Conclusion

This lecture briefly covered the following topics:

- Setting up single trait random regression model
- multiple trait model without trait groups
- multiple trait model using trait groups
- parameter files for these models
- format of solution files



Questions?

