

ASREML Workshop
June 13-14, 2005

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Command File (.as file)

- The command file has three main parts
 1. Data input and transformation
 2. Model specification
 3. Variance specification
- Consider an experiment comparing the oxygen consumption of four species of frog.(Zar (1999))¹

¹Original data was obtained from <http://www.statsci.org/data/general/frogs.html>

Frog example

- Dependent Variable
 - Oxygen Consumption
- Fixed Effects
 - Species (4 levels)
 - Temperature (Low and High)
 - Exercise (Rest and Exercise)
- Random effects
 - Frog (nested w/in Species and Temperature)
 - Error

Data

Subject	Species	Temperature	Rest	Exercise
1	1	Low	0.107	0.152
2	1	Low	0.114	0.163
3	1	High	0.133	0.194
...				

- Rest Resting oxygen consumption
- Exercise Exercise oxygen consumption
- One record/subject
- Tab delimited

Data input

SAS

```
data frogs;
  length cond $ 8;
  infile "h:/IAState/frogs.txt" expandtabs firstobs=2;
  input frog species temp $ rest exercise;
  Y=rest;cond="Rest";output;
  Y=Exercise;cond="Exercise";output;
proc print;
```

ASReml

```
Oxygen consumption of frogs
Frog *
Species * !I
Temp * !A
Rest # Resting oxygen consumption
Exercise # Exercise oxygen consumption
frogs.txt !SKIP 1
```

Model Specification

SAS

```
proc mixed data=frogs;
  class frog species temp Cond;
  model y=species|temp|cond;
  random frog(species temp);
run;
```

ASReml

Model Equation

Rest Exercise ~ mu Trait*Species*Temp !r Frog.Species.Temp

- * Expands into main affects and interactions.
- Trait Automatic variable

Variance structure

```
0 0 1
Frog.Species.Temp 1
0 0 I 1
```

R header Line

0 0 1 or s d g

- s Number of heterogeneous R groups (Usually 1)
Zero in our case says $e \sim N(\mathbf{0}, I\sigma^2)$
- d Number terms in the direct product (Usually 1 or 2)
Zero in our case since we are using the default covariance structure
- g Number of random effects
We have one random effect

Random effects

- Header line Effect parts
 - Frog.Species.Temp 1
 - Effect Frog.Species.Temp
 - 1 one part

- Structure line n 1 M <initial values>
 - n=0 number of levels (Zero tells ASReml to figure it out)
 - l=0
 - M=I Covariance structure (In this case an identity)
 - 1 Initial estimate

Results

Number of Observations Read	32
Number of Observations Used	32
Number of Observations Not Used	0

Iteration History

Iteration	Evaluations	-2 Res Log Like	Criterion
0	1	-102.88571654	
1	1	-108.28703666	0.00000000

Convergence criteria met.

The Mixed Procedure

Covariance Parameter Estimates

Cov Parm	Estimate
frog(species*temp)	0.000033
Residual	0.000014

Fit Statistics

-2 Res Log Likelihood	-108.3
AIC (smaller is better)	-104.3
AICC (smaller is better)	-103.4
BIC (smaller is better)	-102.7

ASReml (.asr file)

1 LogL= 68.4455	S2= 0.20438E-04	16 df	1.000	1.000
2 LogL= 68.6518	S2= 0.18062E-04	16 df	1.000	1.324
3 LogL= 68.8008	S2= 0.15811E-04	16 df	1.000	1.793
4 LogL= 68.8462	S2= 0.14259E-04	16 df	1.000	2.288
5 LogL= 68.8465	S2= 0.14126E-04	16 df	1.000	2.340
Final parameter values			1.0000	2.3407

Source	Model terms	Gamma	Component	Comp/SE	% C
Variance	32 16	1.00000	0.141262E-04	2.00	0 P
Frog.Species.Temp	identity 128	2.34071	0.330653E-04	1.622402	U

Warning: LogL Converged; Parameters Not Converged

Single Trait Animal Model

- Weight data on 386 mice
- Five generations
- Birth date covariate
- Missing data

Data input

- Tab delimited file with a header line

Anim	Sire	Dam	Day	Gen	Gender	2Wk	3Wk
6515	5002	4636	31	0	F	104	150
6696	4531	5306	32	0	F	112	164

- Body Weights collected to the nearest .1 of the gram


```

Single Trait Analysis  Title
  Anin * !P
  Sire * !P
  Dam * !P
  Day  Covariate or Dependent Variable
  Generation * !I  Integer Factor
  Gender * !A  Alpha Factor
  BW2

```

- Title line
- Fields indented by a space
- * Number of levels determined by ASREML

```

BW12
M42.dat # Pedigree file
M42.dat !SKIP 1 !MAXIT 100 !EXTRA 5 #Data file

```

- Not indented
- File name
- !Skip 1 skip header record
- !MAXIT 100 Maximum number of iterations 100
- !EXTRA 5 Do an extra 5 iterations after convergence

Pedigree

- Individual, Sire, Dam
- Individual, Sire, Maternal Grand Sire !MGS
- May be Alphanumeric !ALPHA
- Missing parents coded as zero
- Pedigree fields denoted by !P

Single Trait Analysis

Anim * !P

Sire * !P

Dam * !P

Day

Generation * !A

Gender * !I

BW2

M42.dat # Pedigree file

M42.dat !SKIP 1 !NAXIT 100 !EXTRA 5 #Data file

- Not indented

Transformations

```

BW4_5
BW6  !/10
BW8  !/10
BW12 !/10
ADG  !=V11 !=V10 !=/14
day  !=V4  !=/365
M42.dat # Pedigree file

```

$$BW6/ = 10$$

$$ADG = (BW8 - BW12)/14$$

$$day = DAY/365$$

Model Specification

- Fixed
 - Generation, Gender, and Generation*Gender
 $A*B \equiv A B A.B$
 - Day(Generation) covariate
 Generation/Day
- Random
 - Animal $N(\mathbf{0}, A\sigma_A^2)$

```

M42.dat 13SKIP 1 IMAXIT 100 EXTRA 5 #Data file
BW6 ~ mu Generation*Gender Generation.day !r Anim
0 0 1      # Groups Direct_Prod Random_Effects
Anim 1     # Term Direct_Prod
0 0 AINV 1 # Size Sort Structure Initial_Val

```

- Model equation
- Residual and Number of Random effects
- Random effects

Results

Model term	Size	Type	COL	Minimum	Mean	Maximum	#zero	#miss
1 Anim	419	Factor	1	51	235.2692	419	0	0
2 Sire	309	Factor	2	9	147.2308	309	0	0
3 Dam	324	Factor	3	3	153.6868	324	0	0
4 Day			4	149.0	313.7	479.0	0	0
5 Generation	5	Factor	5	2	3.5220	5	0	0
6 Gender	2	Factor	6	1	1.3736	2	0	0
7 BW2			7	63.00	100.8	124.0	0	0
8 BW3			8	83.00	152.8	201.0	0	0
9 BW4_5			9	216.0	274.4	352.0	0	270
10 BW6	1	Variate	10	14.60	29.97	41.50	0	0
11 BW8			11	11.30	32.40	47.30	0	0
12 BW12			12	24.30	35.34	50.40	0	284
13 ADG			13	0.8957	0.1737	0.7429	1	0
14 day			14	0.4082	0.8596	1.312	0	0
15 mu	1	Constant Term						
16 Generation.Ge	10	Interaction	5	5 Generation:	5	6 Gender	:	2
17 Generation.da	5	Interaction	5	5 Generation:	5	14 day	:	1
419 Ainverse		1.0000						

NOTICE: 11 (more) singularities,

1	LogL=-519.561	S2= 4.2063	352 df	1.000	1.000
2	LogL=-519.429	S2= 4.1229	352 df	1.000	1.055
3	LogL=-519.134	S2= 3.8923	352 df	1.000	1.222
4	LogL=-518.913	S2= 3.6032	352 df	1.000	1.468
5	LogL=-518.854	S2= 3.3588	352 df	1.000	1.713
6	LogL=-518.854	S2= 3.3592	352 df	1.000	1.712
7	LogL=-518.854	S2= 3.3591	352 df	1.000	1.712
8	LogL=-518.854	S2= 3.3592	352 df	1.000	1.712
9	LogL=-518.854	S2= 3.3592	352 df	1.000	1.712
10	LogL=-518.854	S2= 3.3592	352 df	1.000	1.712
11	LogL=-518.854	S2= 3.3592	352 df	1.000	1.712
Final parameter values				1.0000	1.7122

Source	Model	terms	Gamma	Component	Comp/SE	% C
Variance		364 352	1.00000	3.35915	4.28	0 P
Anim	Ainverse	419	1.71225	5.75170	3.84	0 U

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June 13-14, 2005

Analysis of Variance	DF	F-incr		
5 Generation	3	9.77		
6 Gender	1	335.48		
16 Generation.Gender	3	2.20		
17 Generation.day	4	12.26		
	Estimate	Standard Error	T-value	T-prev
17 Generation.day				
2	-12.1514	34.7302	-0.35	
3	-64.1232	82.9213	-0.77	-0.58
4	285.773	41.3023	6.92	3.80
5	-14.1994	43.4579	-0.33	-5.09
16 Generation.Gender				
11	0.501526	0.810515	0.62	
13	-0.698691	0.787780	-0.89	
15	-1.57254	0.778862	-2.02	
6 Gender				
M	5.57881	0.562568	9.92	
5 Generation				
2	38.6446	59.8426	0.65	
3	-290.988	44.0447	-6.61	-4.64
4	11.4335	57.9476	0.20	4.41
15 mu				
23	34.5298	14.5634	2.37	
1 Anim				
		419 effects fitted		
3 possible outliers: see .res file				

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SLOPES FOR LOG(ABS(RES)) on LOG(PV) for Section 1
 -0.15

```

          ** *
          *** **
          *****
          *****
          *****
          *****
          ***** * *
          *****
          *****
          *****
          *****
          *****
          * *****
          **** *****
    *      * * ** ***** *
    
```

Min Mean Max -8.0925 0.34521E-12 3.8418 omitting 0 zeros

STND RES 210 -4.75
 STND RES 295 -6.71
 STND RES 352 -3.76

209	29.789	-0.1889	1.699
210	25.038	-5.738	1.699
211	30.665	0.8347	1.699
...			
294	28.530	-0.4301	1.650
295	22.693	-8.093	1.665
296	28.926	-0.3260	1.665
...			
351	25.686	-0.3862	1.607
352	27.540	-4.540	1.682
353	30.953	-0.2526	1.682

Anim	Sire	Dam	Day	Gen	Gender	2Wk	3Wk	4.5Wk	6Wk	8Wk	12Wk
...											
6323	3962	3313	362	3	M	106	147	.	296	339	.
6324	3962	3313	362	3	M	107	147	.	193	237	.
6325	3962	3313	362	3	M	118	162	.	315	338	.
...											
8778	6971	6523	469	4	F	093	140	.	281	333	.
8981	7172	6207	469	4	M	080	120	.	146	113	.
8982	7172	6207	469	4	M	079	126	.	286	333	.
...											
9638	6202	6526	473	4	F	091	138	.	253	271	.
9641	6813	7188	473	4	M	091	119	.	230	300	.
9642	6813	7188	473	4	M	092	127	.	307	311	.

Heritability (.pin file)

Source	Model	terms	Gamma	Component	Comp/SE
Variance	364	352	1.00000	3.35915	4.28
Anim	Ainverse	419	1.71225	5.75170	3.84

$$\sigma_1^2 = \sigma_E^2 \quad \sigma_2^2 = \sigma_A^2$$

- Linear Functions

$$\sigma_{nvc+1}^2 = \sum_{i=1}^{nvc} \sigma_i^2 \lambda_i$$

- Ratios

$$\theta = \sigma_k^2 / \sigma_l^2$$

mouse_42_st.pin

F Add. 2 $\sigma_3^2 = \sigma_A^2 = \sigma_1^2$
 F Env. 1 $\sigma_4^2 = \sigma_E^2 = \sigma_2^2$
 F Phen. 1 + 2 * 1 $\sigma_5^2 = \sigma_P^2 = \sigma_1^2 + \sigma_2^2 \times 1$
 H h2 2 5 $h^2 = \sigma_2^2 / \sigma_5^2$

- F linear functions of variance components $\sigma_P^2 = \sigma_A^2 + \sigma_E^2 * 1$
- H ratios of variance components $h^2 = \sigma_A^2 / \sigma_P^2$

Results

ASreml -P mouse_42_st

3 Add. 2	5.752	1.498		
4 Env. 1	3.359	0.7848		
5 Phen. 1	9.111	0.9741		
h2	= Anim	2/Phen. 1	5=	0.6313 0.1095

Effect	Variance	Standard Error
Add.	5.752	1.498
Env.	3.359	0.9741
Phen.	9.111	0.9741
h ²	0.6313	0.1095

Correlated Random Effects

- Direct and Maternal Genetic Effects
- Random Regression
- Model Specification ! [Animal Dam !]
- Variance Specification

Direct and Maternal genetic effects

- Fixed
 - Generation, Gender, and Generation*Gender
 - Day(Generation) covariate
- Random
 - Animal Dam $N(\mathbf{0}, G \otimes A)$
 - ide(Dam) $N(\mathbf{0}, I\sigma_L^2)$

BW6 ~ mu Generation*Gender Generation.day,
!r ![Anim Dam !] ide(Dam)

Variance Structures

```

0 0 2
Anim 2
2 0 US 1 .1 1 !GP # $\sigma_1^2$   $\sigma_{12}$   $\sigma_2^2$  Lower Triangle by row
0 0 AINV #No initial value
ide(Dam) 1
0 0 I 1

```

ASReml mouse_42_dm

```

1 LogL=-522.333 S2= 3.5466 352 df
2 LogL=-519.787 S2= 3.4493 352 df : 1 components constrained
Warning: EM updates for 1 positive definite US structure(s).
3 LogL=-514.341 S2= 3.0163 352 df1
4 LogL=-514.167 S2= 2.9896 352 df
5 LogL=-514.164 S2= 2.9944 352 df
6 LogL=-514.158 S2= 3.0088 352 df
7 LogL=-514.151 S2= 3.0301 352 df
8 LogL=-514.148 S2= 3.0561 352 df
9 LogL=-514.148 S2= 3.0661 352 df
10 LogL=-514.148 S2= 3.0694 352 df
11 LogL=-514.148 S2= 3.0705 352 df
12 LogL=-514.148 S2= 3.0709 352 df
13 LogL=-514.148 S2= 3.0710 352 df
14 LogL=-514.148 S2= 3.0710 352 df
Warning: 1 variance structures were modified to make them positive definite
ASREML may have fixed the structure [flagged by B]
and may not have converged to a maximum likelihood solution.
Suggestion: rerun with -C option
or maybe try an ANTE or XFA structure

```

Source	Model	terms	Gamma	Component	Comp/SE	% C
Variance	364	352	1.00000	3.07102	2.21	0 P
Anim	UnStruct	1	1.73603	5.33138	1.91	0 P
Anim	UnStruct	1	-0.482726	-1.48246	-0.83	0 P
Anim	UnStruct	2	0.202433	0.621676	0.32	0 P
ide(Dam)	identity	419	0.425535	1.30683	1.07	0 U

Covariance/Variance/Correlation Matrix UnStructured

1.736	-0.8143	σ_1^2	ρ_{12}
-0.4827	0.2024	σ_{12}	σ_2^2

Multiple Trait Models

- Multiple dependent variables
- Trait automatic variable
- $e \sim N(\mathbf{0}, \otimes R_i)$
- Trait specific models (at())
- Correlations (R c12 v1 v2)

Multiple Dependent Variables

- Dependent Variables

- Body Weight at 6 weeks
- Body Weight at 8 weeks

```

BW6 BW8 ~ Trait*Generation*Gender Trait.Generation.day mv ,
!r ![ Trait.Anim Trait.Dam !]
1 2 1
0 0 I
2 0 US 3.885 3.271 6.963 !GP
Trait.Anim 2
4 0 US 3.724 4.356 6.372 .1 .1 1.927 .1 .1 1.414 1.329 !GP
0 0 AINV

```

Results

Source	Model terms	Gamma	Component	Comp/SE	% C
Variance	772 748	1.00000	1.00000	1.50	0 C
Residual	UnStruct 1	2.50146	2.50146	0.00	0 P
Residual	UnStruct 1	2.12496	2.12496	4.42	0 P
Residual	UnStruct 2	5.39939	5.39939	8.51	0 P
Trait.Anim	UnStruct 1	6.61415	6.61415	0.00	0 P
Trait.Anim	UnStruct 1	6.78293	6.78293	7.67	0 P
Trait.Anim	UnStruct 2	9.55886	9.55886	13.94	0 P
Trait.Anim	UnStruct 1	-2.30962	-2.30962	-7.73	0 P
Trait.Anim	UnStruct 2	-4.00834	-4.00834	-2.06	0 P
Trait.Anim	UnStruct 3	2.66842	2.66842	1.66	0 P
Trait.Anim	UnStruct 1	-1.84080	-1.84080	-1.09	0 P
Trait.Anim	UnStruct 2	-4.29757	-4.29757	-1.90	0 P
Trait.Anim	UnStruct 3	2.66994	2.66994	1.88	0 P
Trait.Anim	UnStruct 4	3.45231	3.45231	1.85	0 P

Warning: Code B - fixed at a boundary (!GP) F - fixed by user
 ? - liable to change from P to B P - positive definite
 C - Constrained by user (!VCC) U - unbounded
 S - Singular Information matrix

S means there is no information in the data for this parameter.

Very small components with Comp/SE ratios of zero sometimes indicate poor

scaling. Consider rescaling the design matrix in such cases.

Covariance/Variance/Correlation Matrix UnStructured

2.501 0.5782
2.125 5.399

Covariance/Variance/Correlation Matrix UnStructured

6.614 0.8531 -0.5498 -0.3852 σ_{a6}^2
6.783 9.559 -0.7937 -0.7481 σ_{a86} σ_{a8}^2
-2.310 -4.008 2.668 0.8797 σ_{m6a6} σ_{m6a8} σ_{m6}^2
-1.841 -4.298 2.670 3.452 σ_{m8a6} σ_{m8a8} σ_{m86} σ_{m8}^2

Analysis of Variance	DF	F-incr
5 Generation	3	10.49
6 Gender	1	363.81
16 Trait.Generation	3	3.31
17 Trait.Gender	1	23.65
18 Generation.Gender	3	1.67
19 Trait.Generation.Gender	3	1.71
21 Trait.Generation.day	8	6.53

- As the models become more complicated the choice of starting values becomes more important
- Simple single trait models
- Complex single trait models
- Simple two trait models
- Complex two trait models

Functions of Variance Components

- Group terms together

- Maternal $\sigma_{m6}^2 = \sigma_{15} \sigma_{m68} = \sigma_{16} \sigma_{m8}^2 = \sigma_{17}$

F M1 10 #15:17

F M12 13

F M2 14

- Direct*Maternal $\sigma_{m6a6} = \sigma_{18} \sigma_{m6a8} = \sigma_{19} \sigma_{m8a6} = \sigma_{20} \sigma_{m8a8} = \sigma_{21}$

F AM1 8 18:20

F AM12 9 *.5 11*.5

F AM22 12

- Phenotypic Variances and Covariances $\sigma_{p6}^2 = \sigma_{21} \sigma_{p68} = \sigma_{22} \sigma_{p8}^2 = \sigma_{23}$

F Phen. 5:7 + 15:17 +18:20 #21:23

- Correlations

$$\frac{\sigma_{xy}}{\sqrt{\sigma_x^2 \sigma_y^2}}$$

- R x xy y

R Dr 5 6 7

R Mr 15 16 17

R Pr 21 22 23

15 M1 10	2.668	1.604		
16 M12 13	2.670	1.421		
17 M2 14	3.452	1.864		
18 AM1 8	-2.310	0.2987		
19 AM12 9	-2.925	1.694		
20 AM22 12	-4.298	2.257		
21 Phen. 5	6.973	2.890		
22 Phen. 6	6.528	0.5667		
23 Phen. 7	8.714	-1.899		
h2D1	= Trait.An 5/Phen. 5 21=		0.9485	0.2649
h2D2	= Trait.An 7/Phen. 7 23=		1.0970	0.1355
Dr	= Trait.An/SQR[Trait.An*Trait.An]=		0.8531	0.1000
Mr	= M12 13 /SQR[M1 10 *M2 14]=		0.8797	0.1962
Pr	= Phen. 6/SQR[Phen. 5*Phen. 7]=		0.8375	0.2254

Trait Specific Models

- Sometimes will want to include an effect only for a subset of traits

– at(f,1)

* 1 when factor f is at level 1

* 0 otherwise

Trait	Trt	Trt	at(Trait,1).Trt
1	A	(1 0 0)	(1 0 0)
1	B	(0 1 0)	(0 1 0)
1	C	(0 0 1)	(0 0 1)
2	A	(1 0 0)	(0 0 0)
2	B	(0 1 0)	(0 0 0)

Model

- Traits
 - 6 week weight
 - ADG between 6 and 8 weeks
- Fixed effects
 - Generation*Gender
 - mv (missing values)
 - day(Generation) covariate

- Random effects
 - Direct and Maternal effects for 6 week weight
 - Direct effect for ADG

$$\begin{pmatrix} \sigma_{a6}^2 & \sigma_{a68} & \sigma_{a6m6} \\ \sigma_{a86} & \sigma_{a8}^2 & \sigma_{a8m6} \\ \sigma_{m6a6} & \sigma_{m6a8} & \sigma_{m6}^2 \end{pmatrix} \otimes A$$

```

DWB ADG ~ Trait*Generation*Gender Trait.Generation.cov
+ r ~ [ Trait.Anim at(Trait,1).Dam () ] if mv
1 1 1
0 0 1
2 0 10 8 000 0.071 0.965 1CF
Trait.Anim >
3 0 00 1 101 2 11 1 0 1CF
> 0 111V

```


Solution file .sln

```

Trait.Anim          1.004636          0.1936          2.440
Trait.Anim          1.005002          0.1936          2.440
...
Trait.Anim          1.009757          -1.659          1.795
Trait.Anim          1.009758          -1.294          1.795
Trait.Anim          2.004636          -0.4696E-01    0.1062
Trait.Anim          2.005002          -0.4696E-01    0.1062
...
Trait.Anim          2.009757          0.5524E-01     0.8915E-01
Trait.Anim          2.009758          0.3535E-01     0.8915E-01
at(Trait,1).Dam    1.004636          0.7488          1.596
at(Trait,1).Dam    1.005002          0.7488          1.596
...
at(Trait,1).Dam    1.009757          -0.1611         1.450
at(Trait,1).Dam    1.009758          -0.1009         1.450

```

ASREML Workshop

June 13-14, 2005

```

1 LogL=-62.3268    S2= 1.0000        748 df
2 LogL=-62.3268    S2= 1.0000        748 df
3 LogL=-62.3268    S2= 1.0000        748 df

Source            Model terms      Gamma    Component    Comp/SE    % C
Variance          772    748    1.00000    1.00000    1.44    0 C
Residual          UnStruct    1    2.71788    2.71788    0.00    0 P
Residual          UnStruct    1    -0.380548E-01    -0.380548E-01    -0.23    0 P
Residual          UnStruct    2    0.197066E-01    0.197066E-01    0.01    0 P
Trait.Anim        UnStruct    1    6.16375    6.16375    0.00    0 P
Trait.Anim        UnStruct    1    0.360471E-01    0.360471E-01    0.06    0 P
Trait.Anim        UnStruct    2    0.122103E-01    0.122103E-01    2.69    0 P
Trait.Anim        UnStruct    1    -2.23893    -2.23893    -1.42    0 P
Trait.Anim        UnStruct    2    -0.117956    -0.117956    -1.82    0 P
Trait.Anim        UnStruct    3    2.73618    2.73618    2.05    0 P
Warning: Code B - fixed at a boundary (!GP)    F - fixed by user
...
Covariance/Variance/Correlation Matrix UnStructured
2.718    -0.1644
-0.3805E-01    0.1971E-01
Covariance/Variance/Correlation Matrix UnStructured
6.164    0.1314    -0.5452
0.3605E-01    0.1221E-01    -0.6453
-2.239    -0.1180    2.736

```

ASREML Workshop

June 13-14, 2005

Random Regression

- General term used when the model includes random covariates
- Common examples
 - Polynomials
 - Often use orthogonal polynomials for numerical reasons
 - Splines
 - Piecewise polynomials
 - Knots where the pieces join

Polynomials

- Arrange data with one record per Weight and create a "Time" covariate

Anim	Sire	Dam	Day	Gen	Gender	Week	Weight
6515	5002	4636	31	0	F	2	104
6515	5002	4636	31	0	F	3	150
6515	5002	4636	31	0	F	4.5	226
6515	5002	4636	31	0	F	6	.
6515	5002	4636	31	0	F	8	.
6515	5002	4636	31	0	F	12	.
6696	4531	5306	32	0	F	2	112
6696	4531	5306	32	0	F	3	164
6696	4531	5306	32	0	F	4.5	222
6696	4531	5306	32	0	F	6	.

Data Specification

```

Two Trait Analysis
Anim 300 !P
Sire 300 !P
Dam 300 !P
Day
Generation * !I
Gender * !A
Week
Weight !/10
day !=V4 !/365
M42.dat # Pedigree file
M42_rr.dat !SKIP 1 !MAXIT 100 #Data file

```

Legendre Polynomials

- $\text{leg}(x,p)$: construct Legendre polynomials of order p
 - Linear transformation of

$$\begin{pmatrix} 1 & x & x^2 & \cdots & x^p \end{pmatrix} \rightarrow \begin{pmatrix} P_0(x) & P_1(x) & P_2(x) & \cdots & P_p(x) \end{pmatrix}$$

- $p < 0$ drop the constant term
- Mathematically equivalent to a standard polynomial
- Advantages are numerical

Model

- Dependent Variable: Weight
- Fixed Effects:
 - Week*Generation*Gender contemporary groups
 - day(Generation*Week) covariates
- Random effects
 - Third order Legendre polynomials direct effects

- Residual
 - Second order Legendre polynomial
 - Heterogeneous Variance

Variance Structure

$$\begin{aligned}
 G &= \begin{pmatrix} P_0(2) & P_1(2) & P_2(2) & P_3(2) \\ P_0(3) & P_1(3) & P_2(3) & P_3(3) \\ & & \vdots & \\ P_0(12) & P_1(12) & P_2(12) & P_3(12) \end{pmatrix} \begin{pmatrix} \sigma_{a0}^2 & \sigma_{a01} & \sigma_{a02} & \sigma_{a03} \\ \sigma_{a10} & \sigma_{a1}^2 & \sigma_{a12} & \sigma_{a13} \\ \sigma_{a20} & \sigma_{a21} & \sigma_{a2}^2 & \sigma_{a23} \\ \sigma_{a30} & \sigma_{a31} & \sigma_{a32} & \sigma_{a3}^2 \end{pmatrix} \begin{pmatrix} P_0(2) & P_0(3) & \dots & P_0(12) \\ P_1(2) & P_1(3) & \dots & P_1(12) \\ P_2(2) & P_2(3) & \dots & P_2(12) \\ P_3(2) & P_3(3) & \dots & P_3(12) \end{pmatrix} \\
 &= \sum_{i=0}^3 \sum_{j=0}^3 [P_i P_j'] \sigma_{aij} \\
 P_j' &= (P_j(2) \quad P_j(3) \quad \dots \quad P_j(12)) \\
 R &= \sum_{i=0}^2 \sum_{j=0}^2 [P_i P_j'] \sigma_{peij} + \text{Diag}(\sigma_{ri}^2)
 \end{aligned}$$

```

Weight ~ fac(Week).Generation*Gender fac(Week).Generation.day
  |r  leg(Week,3).Anim ide(Anim).leg(Week,2) !f mv
1 2 2
386 0 I
6 0 DIAG 1 2 3 4 5 6 !S2==1 !GP
leg(Week,3).Anim 2
4 Week US !GP !+10
16
9 6
0 0 1
0 0 0 .7
419 0 ANIV
ide(Anim).leg(Week,2) 2
419 0 IDEK
6 0 US !GP !+0
1
0 1
0 0 1

```

Source	Model	terms	Gamma	Component	Comp/SE	% C
Variance	2316	1624	1.00000	1.00000	1.19	0 F
Residual	DIAGonal	1	0.100000E-06	0.100000E-06	0.00	0 B
Residual	DIAGonal	2	0.568810	0.568810	0.57	0 P
Residual	DIAGonal	3	1.13715	1.13715	0.83	0 P
Residual	DIAGonal	4	2.35523	2.35523	0.00	0 P
Residual	DIAGonal	5	2.14743	2.14743	9.59	0 P
Residual	DIAGonal	6	1.14547	1.14547	1.32	-1 P
leg(Week,3).Anim	UnStruct	1	7.15231	7.15231	4.92	0 P
leg(Week,3).Anim	UnStruct	1	3.82721	3.82721	2.99	0 P
leg(Week,3).Anim	UnStruct	2	2.39690	2.39690	1.47	0 P
leg(Week,3).Anim	UnStruct	1	-0.467647	-0.467647	-1.09	0 P
leg(Week,3).Anim	UnStruct	2	-0.290318	-0.290318	-1.13	0 P
leg(Week,3).Anim	UnStruct	3	0.166023	0.166023	1.33	0 P
leg(Week,3).Anim	UnStruct	1	-1.10017	-1.10017	-2.80	0 P
leg(Week,3).Anim	UnStruct	2	-0.735352	-0.735352	-2.57	0 P
leg(Week,3).Anim	UnStruct	3	-0.599325E-01	-0.599325E-01	-0.41	0 P
leg(Week,3).Anim	UnStruct	4	0.957631	0.957631	4.27	0 P
ide(Anim).leg(Week,2)UnStruct	1	4.46775	4.46775	3.98	0 P	
ide(Anim).leg(Week,2)UnStruct	1	2.69488	2.69488	3.18	0 P	
ide(Anim).leg(Week,2)UnStruct	2	2.17389	2.17389	3.01	0 P	
ide(Anim).leg(Week,2)UnStruct	1	-0.153679	-0.153679	-0.43	0 P	
ide(Anim).leg(Week,2)UnStruct	2	0.392403E-01	0.392403E-01	0.12	0 P	
ide(Anim).leg(Week,2)UnStruct	3	0.284981	0.284981	1.33	0 P	

Warning: Code B - fixed at a boundary (!GP) F - fixed by user
? - liable to change from P to B P - positive definite

C - Constrained by user (!VCC) U - unbounded
S - Singular Information matrix
S means there is no information in the data for this parameter.
Very small components with Comp/SE ratios of zero sometimes indicate poor scaling. Consider rescaling the design matrix in such cases.

Covariance/Variance/Correlation Matrix UnStructured

7.152	0.9243	-0.4292	-0.4204
3.827	2.397	-0.4602	-0.4854
-0.4676	-0.2903	0.1660	-0.1503
-1.100	-0.7354	-0.5993E-01	0.9576

Covariance/Variance/Correlation Matrix UnStructured

4.468	0.8647	-0.1362
2.695	2.174	0.4985E-01
-0.1537	0.3924E-01	0.2850

.res File

```

leg(Week,3)                has 4 levels
 0.500  0.707 -1.225  1.581 -1.871
 1.000  0.707 -1.118  1.187 -0.998
 1.500  0.707 -1.012  0.828 -0.318
 2.000  0.707 -0.905  0.505  0.186
 2.500  0.707 -0.799  0.218  0.533
 3.000  0.707 -0.692 -0.033  0.742
 3.500  0.707 -0.586 -0.248  0.830
 4.000  0.707 -0.479 -0.427  0.818
 4.500  0.707 -0.373 -0.571  0.722
 5.000  0.707 -0.266 -0.678  0.562
 5.500  0.707 -0.160 -0.750  0.356
 6.000  0.707 -0.053 -0.786  0.122
 6.500  0.707  0.053 -0.786 -0.122
 7.000  0.707  0.160 -0.750 -0.356
 7.500  0.707  0.266 -0.678 -0.562
 8.000  0.707  0.373 -0.571 -0.722
...
11.500  0.707  1.118  1.187  0.998
12.000  0.707  1.225  1.581  1.871

```

ASREML Workshop

June 13-14, 2005

```

leg(Week,2)                has 3 levels
 0.500  0.707 -1.225  1.581
 1.000  0.707 -1.118  1.187
 1.500  0.707 -1.012  0.828
 2.000  0.707 -0.905  0.505
 2.500  0.707 -0.799  0.218
 3.000  0.707 -0.692 -0.033
 3.500  0.707 -0.586 -0.248
 4.000  0.707 -0.479 -0.427
 4.500  0.707 -0.373 -0.571
 5.000  0.707 -0.266 -0.678
 5.500  0.707 -0.160 -0.750
 6.000  0.707 -0.053 -0.786
 6.500  0.707  0.053 -0.786
 7.000  0.707  0.160 -0.750
 7.500  0.707  0.266 -0.678
 8.000  0.707  0.373 -0.571
 8.500  0.707  0.479 -0.427
 9.000  0.707  0.586 -0.248
 9.500  0.707  0.692 -0.033
10.000  0.707  0.799  0.218
10.500  0.707  0.905  0.505
11.000  0.707  1.012  0.828
11.500  0.707  1.118  1.187
12.000  0.707  1.225  1.581

```

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June 13-14, 2005

Covariance Matrix

- .rsv The restart values file is convenient place to grab the variance components.

```

4045 30 985748 4673
  0.000000    0.000000    0.000000    0.000000    0.000000   -0.1000000E-36  -0.1000000E-36   1.0
  0.1000000E-06  0.5688098    1.137154    2.355232    2.147434    1.145474    7.152305    3.8
  2.396900   -0.4676474   -0.2903185    0.1660231   -1.100170   -0.7353516   -0.5993249E-01  0.95
  4.467746    2.694879    2.173887   -0.1536786    0.3924026E-01  0.2849811
RSTRUCTURE          1  2
VARIANCE            1  1  0  1.00000
STRUCTURE           386  0  0
STRUCTURE            6  6  80.100000E-060.568810    1.13715    2.35523
2.14743    1.14547
ide(Anim).leg(Week,2)  2  0
STRUCTURE           419  0  0
STRUCTURE            3  6  9 4.46775    2.69488    2.17389   -.153679
0.392403E-010.284981
...

```

varfunc.sas

```

data leg3;
  input x p0 p1 p2 p3;
  cards;
    0.500    0.707   -1.225    1.581   -1.871
    1.000    0.707   -1.118    1.187   -0.998
    1.500    0.707   -1.012    0.828   -0.318
  ...
    12.000    0.707    1.225    1.581    1.871
data leg2;
  input x p0 p1 p2;
  cards;
    0.500    0.707   -1.225    1.581
    1.000    0.707   -1.118    1.187
  ...
    12.000    0.707    1.225    1.581

```



```

proc iml;

  start vec2var(vector);
    nd=ncol(vector);
    n=(-1+sqrt(1+8*nd))/2;
    matrix=j(n,n,0);
    idx=0;
    do i=1 to n;
      do j=1 to i;
        idx=idx+1;
        matrix[i,j]=vector[idx];
        matrix[j,i]=vector[idx];
      end;
    end;
    return(matrix);
  finish vec2var;

```

```

use leg3;
read all var {p0 p1 p2 p3} into P3;
read all var {x};
use leg2;
read all var {p0 p1 p2} into P2;
print x P3 P2;
xidx=2*{2 3 4.5 6 8 12};
xp=[xidx,];
print tp;

```

```

RC={0.1000000E-06  0.5688098      1.137154
     2.355232      2.147434      1.145474 };
GPC={7.152305      3.827207
     2.396900     -0.4676474     -0.2903185     0.1660231
     -1.100170     -0.7353516     -0.5993249E-01  0.9576310};
RPC={4.467746      2.694879      2.173887      -0.1536786
     0.3924026E-01  0.2849811};
RD=diag(RC);
GP=vec2var(GPC);
RP=vec2var(rpc);
GCov=P3[xidx,]*GP*P3[xidx,]';
RCov=P2[xidx,]*Rp*P2[xidx,]'+Rd;
print Gcov RCov;
quit;
run;

```

Splines

- Smooth piecewise cubic polynomials which join at knots
- Knots at endpoints (3 knots at 2, 8, and 12 weeks)

```
!SPLINE spl(Week,3) 2 8 12
```

- `spl(x,k)`: construct spline with k knots
 - $k-2$ terms

$$Z_i = (Z_{i3} \ \cdots \ Z_{ik})$$

- need to include the $a + b x$ linear component

$$(1 \ X_i)$$

```

M42_rr.dat !SKIP 1 !MAXIT 100 #Data file
!SPLINE spl(Week,3) 2 8 12
Weight ~ fac(Week).Generation*Gender fac(Week).Generation.day ,
!r ![ Anim Week.Anim !] spl(Week,3).Anim ![ ide(Anim) Week.ide(Anim) !],
spl(Week,3).ide(Anim) !f mv
1 2 4
386 0 IDEN
6 0 DIAG 1 2 3 4 5 6 !S2==1 !GP
Anim 2
2 Week US !GP !+3
16
9 6
419 0 AINV
spl(Week,3).Anim 1
0 0 AINV 1 !GP
ide(Anim) 2
2 Week US !GP !+3
1
0 1
419 0 I
spl(Week,3).ide(Anim) 1
0 0 I 1

```

Variance Function

$$\mathbf{G} = \begin{pmatrix} 1 & 2 \\ 1 & 3 \\ \vdots & \\ 1 & 12 \end{pmatrix} \begin{pmatrix} \sigma_{a1}^2 & \sigma_{a12} \\ \sigma_{a21} & \sigma_{a2}^2 \end{pmatrix} + \mathbf{Z} \begin{pmatrix} \sigma_{as}^2 & & & \\ & \sigma_{as}^2 & & \\ & & \dots & \\ & & & \sigma_{as}^2 \end{pmatrix} \mathbf{Z}'$$

Genetic Merit

$$\mathbf{a}_i = 1a_{i1} + \mathbf{x}a_{i2} + \mathbf{Z} \begin{pmatrix} a_{i3} \\ \vdots \\ a_{ik} \end{pmatrix}$$

- .res contains the \mathbf{Z} and \mathbf{x} values
- .sln contains the solutions

predfun.sas

```

data spl3;
  int=1;
  input x spl;
  cards;
0.500    2.547
1.000    2.026
1.500    1.506
....
11.500   0.831
12.000   1.477

data sln int slope spline;
length level $ 15;
length factor $20;
infile "h:/Mouse/mouse_42_spline.sln" expandtabs;
input factor $ level $ sol se;
output sln;
if factor="Anim" then output int;
if factor="Week.Anim" then output slope;
if factor="spl(Week,3).Anim" then output spline;
run;

```

```

proc iml;
  use spl3;
  read all into Z;
  use sln;
  read all;
  use int;
  read all var sol into aint;
  read all var level into Anim;
  use slope;
  read all var sol into aslope;
  use spline;
  read all var sol into aspline;
  GM=aint||aslope||aspline;
  Zx=Z[2 3 4.5 6 8 12*2,];
  print Zx;
  GW=GM*Zx';
  week=W2 W3 W4_5 W6 W8 W12;
  print GW[rowname=Anim colname=week];
  create Gdata from GW[colname=week];
  append from GW;;
  create Anim var Anim;
  append;
  quit;
run;

```

```

data Gdata;
  merge Anim Gdata;
run;

```

	GW					
	W2	W3	W4_5	W6	W8	W12
4636	0.1474189	0.0339654	-0.11938	-0.229476	-0.263939	0.077499
5002	0.1474189	0.0339654	-0.11938	-0.229476	-0.263939	0.077499
6515	0.2948575	0.06794	-0.238763	-0.45896	-0.527868	0.1551275
5306	-0.088076	-0.282496	-0.561779	-0.809342	-1.057036	-1.251446
4531	-0.088076	-0.282496	-0.561779	-0.809342	-1.057036	-1.251446
6696	-0.176172	-0.564971	-1.123482	-1.61857	-2.113953	-2.50293
4486	0.0914031	0.2157538	0.3876952	0.5221701	0.6041539	0.4126174