


PROC BGLIMM: THE SMOOTH TRANSITION TO BAYESIAN ANALYSIS

Danny Modlin



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Greetings everyone. I have been a Training Consultant at SAS since April 2011. Before SAS, I was a teacher in middle school, high school, as well as Teaching Assistant at the University of North Carolina at Wilmington and North Carolina State University. I have a Bachelors of Science in Mathematics from Elon College, a Masters of Mathematics from UNCW, and a Masters of Statistics from NCSU.

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Mixed Modeling Procedures in SAS

Classical		Bayesian
GENMOD	MIXED	GENMOD w/BAYES Statement
GLIMMIX		BGLIMM
NLMIXED		MCMC

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GENERAL LINEAR MIXED MODEL

$$y = X\beta + Z\gamma + \epsilon$$

where y is the vector of observed responses.
 X is the design matrix of predictor variables.
 β is the vector of regression parameters.
 Z is the design matrix of random variables.
 γ is the vector of random-effect parameters.
 ϵ is the vector of random errors.

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TOY EXAMPLE

toy 1 adhesives breaking strength

toy 2

toy 7

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PROC MIXED PROGRAM

```
proc mixed data=sasuser.toy;  
  class adhesive toy;  
  model pressure=adhesive / solution ddfm=kr;  
  random toy;  
run;
```


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PROC MCMC PROGRAM


```
proc mcmc data=toy seed=27513 diag=all dic outpost=mixed
  propcov=quanew thin=25 nbi=5000 ntu=5000 nmc=500000
  plots(smooth)=all mchistory=brief stats=all;
array beta[3];
parms beta: 0;
parms s2t 1;
parms s2g 1;
prior beta: ~ normal(0, var = 1e5);
prior s2: ~ igamma(2.001, scale = 1.001);
random gamma ~ normal(0,var=s2g) subject=toy
  monitor=(gamma) namesuffix=position;
mu = beta[adhesivebeta] + gamma;
model pressure ~ normal(mu, var = s2t);
title "Bayesian Analysis of the Toy Data Set";
run;
```



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GENERALIZED LINEAR MIXED MODELS (GzLMs)

- GzLMs enable modeling random effects and correlated errors for nonnormal data.
 - A linear predictor can contain random effects.
$$\eta = X\beta + Z\gamma$$
 - The random effects are normally distributed.
 - The conditional mean, $\mu|\gamma$, relates to the linear predictor through a link function.
$$g(\mu|\gamma) = \eta$$
 - The conditional distribution (given γ) of the data belongs to the exponential family of distributions.



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GzLMM FORMULATION AND PROC GLIMMIX

$$g(\mu | \gamma) = X\beta + Z\gamma + \varepsilon$$


LINK= option MODEL statement RANDOM statement

$Y|\gamma \sim \text{exponential family}$ — DIST= option

$\text{var}(\gamma) = G$ ← Options in the RANDOM statement

$\text{var}(y|\gamma) = A_{\mu} R A_{\mu}$

RANDOM _RESIDUAL_ statement *REPEATED in BGLIMM



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Features of BGLIMM

- Suite of covariance structures (for both G- and R-side)
- Covariance heterogeneity modeling
- Built-in priors
- Model Comparison via DIC statistic
- Multi-threading of optimal sampling

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PROC BGLIMM Syntax

- | | |
|-------------|--|
| • DATA= | names the input data set |
| • SEED= | random seed for simulation |
| • OUTPOST= | output a data set to contain posterior samples |
| • NBI= | number of burn-in iterations |
| • NMC= | number of Markov chain iterations |
| • NTHREADS= | number of CPUs to run simulations simultaneously |
| • STATS= | posterior statistics |
| • DIAG= | convergence diagnostics |
| • PLOTS= | plotting |

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PROC BGLIMM Syntax

MODEL response = fixed effects / dist= link= ...;

- | | |
|-----------------------------|------------------------|
| • 9 response distributions: | |
| • Binomial | - Negative binomial |
| • Exponential | - Normal |
| • Gamma | - Poisson |
| • Geometric | - Binary |
| • Inverse Gaussian | |
| • 8 link functions: | |
| • Log | - Identity |
| • Logit | - Loglog |
| • Probit | - Complementary loglog |
| • Inverse | - PowerMinus2 |

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PROC BGLIMM Syntax

RANDOM random-effects / sub= group= type= ...;

- SUB= option to identify the subjects for the random effects
- GROUP= option to identify groups by which to vary the covariance parameters; each new level of the grouping effect produces a new set of covariance parameters
- TYPE= option to define the covariance structure of G
 - 13 choices: AR, ARMA, CS, TOEP, UN, VC, ...
- *Multiple RANDOM statements can be used.

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PROC BGLIMM Syntax

REPEATED repeated-effect / sub= group= type= ...;

- A repeated-effect is required to define the proper location of the repeated responses.
- SUB= option to group repeated measures together for the same subject
- GROUP= option to identify groups by which to vary the covariance parameters
- TYPE= option to define the covariance structure
 - 13 choices: AR, ARMA, CS, TOEP, UN, VC, ...

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Sampling


- PROC BGLIMM updates parameters conditionally and sequentially through Gibbs sampling.
 - The fixed-effect parameters are drawn together first at each iteration.
 - The random-effect parameters are updated by subjects.
 - The G-side covariance parameters are then sampled.
 - Lastly, the R-side covariance parameters are updated.
- If present, missing response values are treated as parameters and are thus sampled as well.

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Prior Distributions


- Fixed-effect parameters (Betas)
 - Flat/constant; normal
- Scale parameter
 - Inverse gamma; gamma; improper
- G-side Covariance parameters
 - Inverse wishart; inverse gamma; uniform; halfcauchy; halfnormal; siwishart
- R-side Covariance parameters
 - Inverse wishart; inverse gamma

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Examples using BGLIMM

This demonstration illustrates the concepts discussed previously.



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THANKS FOR ATTENDING. QUESTIONS?



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