LME and Creatinine

Joining Data

```
In [1]: library(data.table)
    library(nlme)
    library(parallel)
    library(xtable)

    creatinine.data <- data.table(read.csv('creat.csv'), key='group')
    head(creatinine.data)</pre>
```

```
        id
        group
        age
        scr

        1
        1
        35.765
        0.182

        1
        1
        37.990
        0.088

        3
        1
        51.083
        0.156

        3
        1
        52.386
        0.116

        3
        1
        52.805
        0.087

        3
        1
        52.997
        0.067
```

group	kidney.disease	hypertension
1	1	1
2	1	0
3	0	1
4	0	0

```
In [3]: creatinine.group.data <- creatinine.data[group.data]
    setkey(creatinine.group.data, id)
    head(creatinine.group.data)</pre>
```

id	group	age	scr	kidney.disease	hypertension
1	1	35.765	0.182	1	1
1	1	37.990	0.088	1	1
2	2	24.997	1.429	1	0
2	2	27.441	1.111	1	0
2	2	30.524	1.429	1	0
3	1	51.083	0.156	1	1

Fitting LME Model

```
In [4]: fit.scr.model <- function(data, interaction.term=FALSE) {
    formula = scr ~ age + kidney.disease
    if (interaction.term) { formula <- update(formula, . ~ . + age:kidney.disease)
}
lme(formula,
    random=reStruct(~age|id, pdClass='pdDiag'),
    correlation=corExp(form=~age|id, nugget=TRUE),
    method='ML', data=data,
    control=lmeControl(maxIter=100, msMaxIter=100, niterEM=50))
}</pre>
```

Covariance for Subject i

```
In [5]: make.covariance <- function(model, i) {
    # Error term, usually denoted epsilon
    error.correlation <- corMatrix(model$model$truct$corStruct)[[as.character(i)]]
    if (is.null(error.correlation)) { error.correlation <- 1 }
    # Random effects correlation, Z * G * tranpose(Z).
    Z <- cbind(1, model$data[J(i), age])
    random.correlation <- Z %*% as.matrix(model$model$truct$reStruct$id) %*% t(Z)
    # Convert correlation matrix into covariance matrix.
    (random.correlation + error.correlation)*(model$sigma*model$sigma)
}</pre>
```

β_i Covariance

Maximum Likelihood Estimate

This assumes that the random effects model is correct. I can also be used as the bread part of the sandwich estimator.

```
In [6]: make.covariates <- function(model, data) model.matrix(model$terms, data)

make.ml.parameter.covariance <- function(model) {
    groups <- unique(model$groups)$id
    chol2inv(chol(Reduce(`+`, lapply(groups, function(i) {
        X <- make.covariates(model, model$data[J(i)])
        t(X) %*% chol2inv(chol(make.covariance(model, i))) %*% X
    }))))
}</pre>
```

Sandwich Estimate

```
In [7]: make.response <- function(data) data$scr

make.sandwich.parameter.covariance <- function(model) {
    bread <- make.ml.parameter.covariance(model)
    meat <- Reduce(`+`, lapply(unique(model$groups)$id, function(i) {
        X <- make.covariates(model, model$data[J(i)])
        y <- make.response(model$data[J(i)])
        weights <- chol2inv(chol(make.covariance(model, i)))
        residuals <- as.numeric(make.response(model$data[J(i)]) - X %*% model$coef
    ficients$fixed)
        empirical.covariance <- outer(residuals, residuals)
        t(X) %*% weights %*% empirical.covariance %*% weights %*% X
    }))
    bread %*% meat %*% bread
}</pre>
```

Bootstrap Estimate

To account for clustering of the data, we resample clusters.

Models

Without Interaction Term

```
In [9]: | scr.model <- fit.scr.model(creatinine.group.data)</pre>
         summary(scr.model, adjustSigma=FALSE)
         Linear mixed-effects model fit by maximum likelihood
          Data: data
                 AIC
                          BIC
                                  logLik
           -53.94986 -11.00314 34.97493
         Random effects:
          Formula: ~age | id
          Structure: Diagonal
                                   age Residual
                 (Intercept)
         StdDev: 0.04643211 0.00522239 0.2633414
         Correlation Structure: Exponential spatial correlation
          Formula: ~age | id
          Parameter estimate(s):
             range
                     nugget
         7.8894707 0.1759323
         Fixed effects: list(formula)
                             Value Std.Error DF t-value p-value
         (Intercept)
                        1.5322224 0.03955663 965 38.73490
                        -0.0129154 0.00096383 965 -13.40006
         kidney.disease -0.2819162 0.02626024 617 -10.73548
                                                                   0
          Correlation:
                        (Intr) age
                        -0.849
         age
         kidney.disease -0.352 -0.082
         Standardized Within-Group Residuals:
                 Min
                              Q1
                                        Med
                                                                  Max
                                                       Q3
         -2.49081741 -0.56731338 -0.07887231 0.43113427 4.95897388
         Number of Observations: 1585
         Number of Groups: 619
In [10]: | ml.parameter.covariance <- make.ml.parameter.covariance(scr.model)</pre>
         sandwich.parameter.covariance <- make.sandwich.parameter.covariance(scr.model)</pre>
         bootstrap.parameter.samples <- do.call(rbind, mclapply(</pre>
             replicate(1024, creatinine.group.data, simplify=FALSE), function(data) {
                 resampled.data <- resample.clusters(data)</pre>
                 fit.scr.model(resampled.data)$coefficients$fixed
             }, mc.cores=4))
         bootstrap.parameter.covariance <- cov(bootstrap.parameter.samples)</pre>
```

	Estimate	ML Standard Error	Sandwich Standard Error	Bootstrap Standard Error
(Intercept)	1.53222240	0.0395566329	0.0393479582	0.041530790
age	-0.01291536	0.0009638288	0.0009409584	0.001030028
kidney.disease	-0.28191623	0.0262602392	0.0250273809	0.034516098

With Interaction Term

```
In [12]: scr.model.interaction <- fit.scr.model(creatinine.group.data, interaction.term=TRU
         summary(scr.model.interaction, adjustSigma=FALSE)
         Linear mixed-effects model fit by maximum likelihood
          Data: data
                 AIC
                           BIC
                                 logLik
           -127.9923 -79.67728 72.99617
         Random effects:
          Formula: ~age | id
          Structure: Diagonal
                 (Intercept)
                                     age Residual
                   0.1317468 0.004823202 0.225834
         StdDev:
         Correlation Structure: Exponential spatial correlation
          Formula: ~age | id
          Parameter estimate(s):
             range
                      nugget
         4.6700641 0.2299764
         Fixed effects: list(formula)
                                 Value Std.Error DF t-value p-value
                             1.1906763 0.05423200 964 21.955237 0.0000
         (Intercept)
                            -0.0031083 0.00143568 964 -2.165016 0.0306
         age
                             0.3135748 0.07137345 617 4.393438 0.0000
         kidney.disease
         age:kidney.disease -0.0166492 0.00185709 964 -8.965190 0.0000
          Correlation:
                                          kdny.d
                            (Intr) age
         age
                            -0.928
         kidney.disease
                            -0.760 0.705
         age:kidney.disease 0.718 -0.773 -0.935
         Standardized Within-Group Residuals:
                              Q1
                                         Med
                                                      03
         -2.30553348 -0.53490304 -0.07976542 0.36022352 5.46804808
         Number of Observations: 1585
         Number of Groups: 619
In [13]: ml.parameter.covariance.interaction <- make.ml.parameter.covariance(scr.model.inte
         sandwich.parameter.covariance.interaction <- make.sandwich.parameter.covariance(sc</pre>
         r.model.interaction)
         bootstrap.parameter.samples.interaction <- do.call(rbind, mclapply(</pre>
             replicate(1024, creatinine.group.data, simplify=FALSE), function(data) {
                 resampled.data <- resample.clusters(data)</pre>
                 fit.scr.model(resampled.data, interaction.term=TRUE)$coefficients$fixed
             }, mc.cores=4))
         bootstrap.parameter.covariance.interaction <- cov(bootstrap.parameter.samples.inte
         raction)
```

```
In [14]: standard.errors.interaction <- data.frame(</pre>
              `Estimate`=scr.model.interaction$coefficients$fixed,
              `ML Standard Error`=sqrt(diag(ml.parameter.covariance.interaction)),
              `Sandwich Standard Error`=sqrt(diag(sandwich.parameter.covariance.interaction
              \verb|`Bootstrap Standard Error` = & \texttt{sqrt}( \\ \texttt{diag}( \texttt{bootstrap.parameter.covariance.interactio})|
          n)),
              check.names=FALSE)
          print(xtable(standard.errors.interaction,
                        caption='Standard error estimates for fixed effect parameters with in
          teraction term.',
                        label='tab:standard_errors_interaction',
                        digits=c(0, 6, 6, 6, 6)),
                booktabs=TRUE, file='standard_errors_interaction.tex',
                sanitize.colnames.function=identity,
                sanitize.rownames.function=identity,
                size='small')
          standard.errors.interaction
```

	Estimate	ML Standard Error	Sandwich Standard Error	Bootstrap Standard Error
(Intercept)	1.190676306	0.054231995	0.050269334	0.050769991
age	-0.003108265	0.001435677	0.001193725	0.001216667
kidney.disease	0.313574816	0.071373449	0.068324860	0.070047199
age:kidney.disease	-0.016649153	0.001857089	0.001705009	0.001706189