

Assignment 7: GLMs week 2 (Linear Regression and beyond)

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OVERVIEW

This exercise accompanies the lessons in Environmental Data Analytics on generalized linear models.

Directions

1. Change “Student Name” on line 3 (above) with your name.
2. Work through the steps, **creating code and output** that fulfill each instruction.
3. Be sure to **answer the questions** in this assignment document.
4. When you have completed the assignment, **Knit** the text and code into a single PDF file.
5. After Knitting, submit the completed exercise (PDF file) to the dropbox in Sakai. Add your last name into the file name (e.g., “Salk_A06_GLMs_Week1.Rmd”) prior to submission.

The completed exercise is due on Tuesday, February 25 at 1:00 pm.

Set up your session

1. Set up your session. Check your working directory, load the tidyverse, nlme, and piecewiseSEM packages, import the *raw* NTL-LTER raw data file for chemistry/physics, and import the processed litter dataset. You will not work with dates, so no need to format your date columns this time.
2. Build a ggplot theme and set it as your default theme.

```
#1
getwd()

## [1] "C:/Users/senam/Box Sync/My Documents/MEM classes/Duke Spring 2020/DataAnalytics/Environmental_D

library(tidyverse)

## -- Attaching packages ----- tidyverse 1.3.0 --
## v ggplot2 3.2.1      v purrr   0.3.3
## v tibble  2.1.3      v dplyr  0.8.3
## v tidyr   1.0.0      v stringr 1.4.0
## v readr   1.3.1      v forcats 0.4.0

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()    masks stats::lag()

library(nlme)

##
## Attaching package: 'nlme'

## The following object is masked from 'package:dplyr':
##
## collapse
```

```
library(piecewiseSEM)

## Registered S3 methods overwritten by 'lme4':
##   method                      from
##   cooks.distance.influence.merMod car
##   influence.merMod             car
##   dfbeta.influence.merMod      car
##   dfbetas.influence.merMod     car

##
##   This is piecewiseSEM version 2.1.0.
##
##
##   Questions or bugs can be addressed to <LefcheckJ@si.edu>.

ntl_lter.data <- read.csv("../Data/Raw/NTL-LTER_Lake_ChemistryPhysics_Raw.csv")
litter.data <- read.csv("../Data/Processed/NEON_NIWO_Litter_mass_trap_Processed.csv")

#2
my.theme <- theme_minimal()+
  theme(legend.position = "top")
theme_set(my.theme)
```

NTL-LTER test

Research question: What is the best set of predictors for lake temperatures in July across the monitoring period at the North Temperate Lakes LTER?

3. Wrangle your NTL-LTER dataset with a pipe function so that it contains only the following criteria:

- Only dates in July (hint: use the daynum column). No need to consider leap years.
- Only the columns: lakename, year4, daynum, depth, temperature_C
- Only complete cases (i.e., remove NAs)

4. Run an AIC to determine what set of explanatory variables (year4, daynum, depth) is best suited to predict temperature. Run a multiple regression on the recommended set of variables.

```
#3
ntl_lter.july <- ntl_lter.data %>%
  filter(daynum >=182 & daynum <=212)%>%
  select(lakename:daynum, depth:temperature_C)%>%
  na.exclude()

#4
ntl_AIC <- lm(temperature_C ~ lakename + year4 + daynum + depth, data = ntl_lter.july)
step(ntl_AIC)
```

```
## Start: AIC=24461.34
## temperature_C ~ lakename + year4 + daynum + depth
##
##           Df Sum of Sq    RSS   AIC
## <none>                 120062 24461
## - year4           1      184 120245 24474
## - daynum          1     1346 121407 24568
## - lakename        8     21056 141118 26016
## - depth           1    403139 523201 38770
```

```
##
## Call:
## lm(formula = temperature_C ~ lakename + year4 + daynum + depth,
##     data = ntl_lter.july)
##
## Coefficients:
##             (Intercept)      lakenameCrampton Lake      lakenameEast Long Lake
##                45.17306                4.71362                -1.46041
## lakenameHummingbird Lake      lakenamePaul Lake      lakenamePeter Lake
##                -4.73042                0.99422                1.44048
##      lakenameTuesday Lake      lakenameWard Lake      lakenameWest Long Lake
##                -1.38445                -0.46590                -0.16847
##                year4                daynum                depth
##                -0.01588                0.04157                -1.96540

# results of AIC recommend keeping all variables in the model
summary(ntl_AIC)
```

```
##
## Call:
## lm(formula = temperature_C ~ lakename + year4 + daynum + depth,
##     data = ntl_lter.july)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -7.8938 -3.0274 -0.2114  2.7781 15.2926
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    45.173063   8.248578   5.476 4.45e-08 ***
## lakenameCrampton Lake    4.713617   0.382185  12.333 < 2e-16 ***
## lakenameEast Long Lake   -1.460406   0.343271  -4.254 2.12e-05 ***
## lakenameHummingbird Lake -4.730421   0.459795 -10.288 < 2e-16 ***
## lakenamePaul Lake        0.994222   0.331643   2.998 0.002726 **
## lakenamePeter Lake       1.440479   0.331406   4.347 1.40e-05 ***
## lakenameTuesday Lake    -1.384450   0.336476  -4.115 3.91e-05 ***
## lakenameWard Lake       -0.465900   0.464619  -1.003 0.316003
## lakenameWest Long Lake  -0.168474   0.341961  -0.493 0.622257
## year4              -0.015885   0.004118  -3.857 0.000115 ***
## daynum              0.041574   0.003985  10.432 < 2e-16 ***
## depth             -1.965403   0.010885 -180.566 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.516 on 9710 degrees of freedom
## Multiple R-squared:  0.7803, Adjusted R-squared:  0.78
## F-statistic: 3135 on 11 and 9710 DF, p-value: < 2.2e-16
```

5. What is the final set of explanatory variables that predict temperature from your multiple regression? How much of the observed variance does this model explain?

Answer: Lake, year, day of the month, and depth are all significant predictors of lake temperatures in July. This model explains 78.0% of the variation in the data.

6. Run an interaction effects ANCOVA to predict temperature based on depth and lakename from the same wrangled dataset.

```
#6
ntl_intancova <- lm(data = ntl_lter.july, temperature_C ~ depth * lakename)
summary(ntl_intancova)
```

```
##
## Call:
## lm(formula = temperature_C ~ depth * lakename, data = ntl_lter.july)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -7.6455 -2.9133 -0.2879  2.7567 16.3606
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    22.9455     0.5861  39.147 < 2e-16 ***
## depth         -2.5820     0.2411 -10.711 < 2e-16 ***
## lakenameCrampton Lake    2.2173     0.6804   3.259  0.00112 **
## lakenameEast Long Lake  -4.3884     0.6191  -7.089 1.45e-12 ***
## lakenameHummingbird Lake -2.4126     0.8379  -2.879  0.00399 **
## lakenamePaul Lake       0.6105     0.5983   1.020  0.30754
## lakenamePeter Lake      0.2998     0.5970   0.502  0.61552
## lakenameTuesday Lake   -2.8932     0.6060  -4.774 1.83e-06 ***
## lakenameWard Lake       2.4180     0.8434   2.867  0.00415 **
## lakenameWest Long Lake  -2.4663     0.6168  -3.999 6.42e-05 ***
## depth:lakenameCrampton Lake  0.8058     0.2465   3.268  0.00109 **
## depth:lakenameEast Long Lake  0.9465     0.2433   3.891  0.00010 ***
## depth:lakenameHummingbird Lake -0.6026     0.2919  -2.064  0.03903 *
## depth:lakenamePaul Lake    0.4022     0.2421   1.662  0.09664 .
## depth:lakenamePeter Lake   0.5799     0.2418   2.398  0.01649 *
## depth:lakenameTuesday Lake  0.6605     0.2426   2.723  0.00648 **
## depth:lakenameWard Lake   -0.6930     0.2862  -2.421  0.01548 *
## depth:lakenameWest Long Lake  0.8154     0.2431   3.354  0.00080 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.471 on 9704 degrees of freedom
## Multiple R-squared:  0.7861, Adjusted R-squared:  0.7857
## F-statistic: 2097 on 17 and 9704 DF, p-value: < 2.2e-16
```

7. Is there a significant interaction between depth and lakename? How much variance in the temperature observations does this explain?

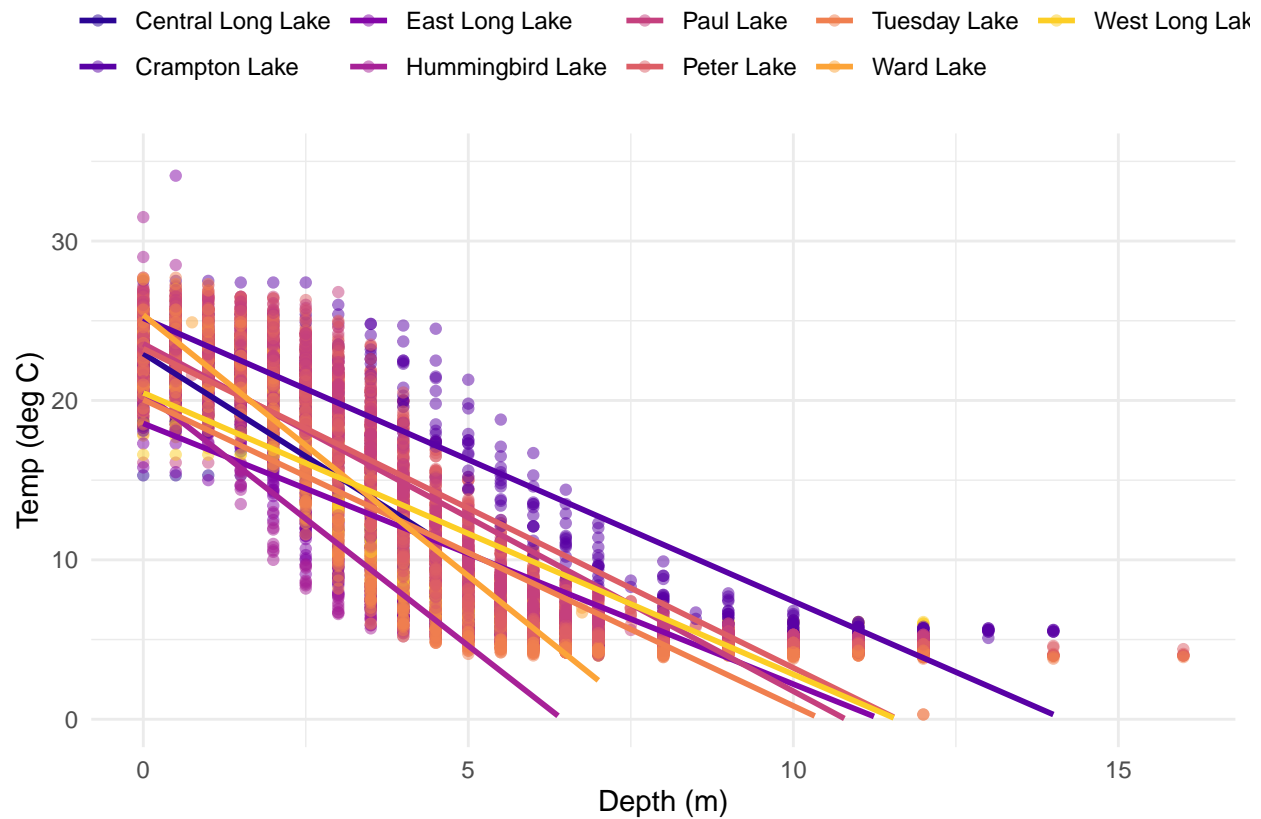
Answer: Yes, there is a significant interaction between depth and lake. This interaction explains 78.6 % of the variation in the data.

8. Create a graph that depicts temperature by depth, with a separate color for each lake. Add a `geom_smooth` (method = "lm", se = FALSE) for each lake. Make your points 50 % transparent. Adjust your y axis limits to go from 0 to 35 degrees. Clean up your graph to make it pretty.

```
#8
ntl_july_fig1 <- ggplot(data = ntl_lter.july, aes(x = depth, y = temperature_C))+
  geom_point(aes(color = lakename), alpha = 0.5)+
  scale_color_viridis_d(option = "plasma", begin = .05, end = .9)+
  labs(x = "Depth (m)", y = "Temp (deg C)", color = "")+
  ylim(0,35)+
  geom_smooth(aes(color= lakename), method = "lm", se = FALSE)
```

```
print(ntl_july_fig1)
```

```
## Warning: Removed 73 rows containing missing values (geom_smooth).
```



9. Run a mixed effects model to predict dry mass of litter. We already know that `nlcdClass` and `functionalGroup` have a significant interaction, so we will specify those two variables as fixed effects with an interaction. We also know that litter mass varies across plot ID, but we are less interested in the actual effect of the plot itself but rather in accounting for the variance among plots. Plot ID will be our random effect.

- Build and run a mixed effects model.
- Check the difference between the marginal and conditional R² of the model.

```
litter.mixed <- lme(data = litter.data, dryMass ~ nlcdClass * functionalGroup, random = ~1|plotID )
summary(litter.mixed)
```

```
## Linear mixed-effects model fit by REML
## Data: litter.data
##      AIC      BIC    logLik
##  9038.575 9179.479 -4493.287
##
## Random effects:
## Formula: ~1 | plotID
##      (Intercept) Residual
## StdDev:   0.5899105 3.456817
##
## Fixed effects: dryMass ~ nlcdClass * functionalGroup
```

	Value	Std.Error
## (Intercept)	0.155492	0.4863580
## nlcdClassgrasslandHerbaceous	-0.156004	0.7789816
## nlcdClassshrubScrub	-0.107080	0.6636775
## functionalGroupLeaves	-0.126008	0.5501061
## functionalGroupMixed	1.477797	0.6323043
## functionalGroupNeedles	7.284064	0.5313161
## functionalGroupOther	-0.048525	0.5500878
## functionalGroupSeeds	-0.058702	0.5501061
## functionalGroupTwigs/branches	1.929441	0.5385556
## functionalGroupWoody material	1.068772	0.5259330
## nlcdClassgrasslandHerbaceous:functionalGroupLeaves	0.181416	0.8847246
## nlcdClassshrubScrub:functionalGroupLeaves	0.173857	0.7510320
## nlcdClassgrasslandHerbaceous:functionalGroupMixed	-0.467648	1.1201304
## nlcdClassshrubScrub:functionalGroupMixed	0.633876	0.9217911
## nlcdClassgrasslandHerbaceous:functionalGroupNeedles	-2.118299	0.8705440
## nlcdClassshrubScrub:functionalGroupNeedles	-2.909142	0.7347172
## nlcdClassgrasslandHerbaceous:functionalGroupOther	0.143603	0.8976715
## nlcdClassshrubScrub:functionalGroupOther	0.104935	0.7528434
## nlcdClassgrasslandHerbaceous:functionalGroupSeeds	0.049290	0.8976827
## nlcdClassshrubScrub:functionalGroupSeeds	0.076708	0.7547591
## nlcdClassgrasslandHerbaceous:functionalGroupTwigs/branches	-0.986627	0.8850639
## nlcdClassshrubScrub:functionalGroupTwigs/branches	-1.503446	0.7409024
## nlcdClassgrasslandHerbaceous:functionalGroupWoody material	-1.017803	0.8802289
## nlcdClassshrubScrub:functionalGroupWoody material	-0.979078	0.7317033
##	DF	t-value
## (Intercept)	1659	0.319706
## nlcdClassgrasslandHerbaceous	9	-0.200266
## nlcdClassshrubScrub	9	-0.161343
## functionalGroupLeaves	1659	-0.229061
## functionalGroupMixed	1659	2.337160
## functionalGroupNeedles	1659	13.709474
## functionalGroupOther	1659	-0.088213
## functionalGroupSeeds	1659	-0.106711
## functionalGroupTwigs/branches	1659	3.582622
## functionalGroupWoody material	1659	2.032144
## nlcdClassgrasslandHerbaceous:functionalGroupLeaves	1659	0.205053
## nlcdClassshrubScrub:functionalGroupLeaves	1659	0.231490
## nlcdClassgrasslandHerbaceous:functionalGroupMixed	1659	-0.417495
## nlcdClassshrubScrub:functionalGroupMixed	1659	0.687657
## nlcdClassgrasslandHerbaceous:functionalGroupNeedles	1659	-2.433305
## nlcdClassshrubScrub:functionalGroupNeedles	1659	-3.959540
## nlcdClassgrasslandHerbaceous:functionalGroupOther	1659	0.159972
## nlcdClassshrubScrub:functionalGroupOther	1659	0.139385
## nlcdClassgrasslandHerbaceous:functionalGroupSeeds	1659	0.054908
## nlcdClassshrubScrub:functionalGroupSeeds	1659	0.101632
## nlcdClassgrasslandHerbaceous:functionalGroupTwigs/branches	1659	-1.114752
## nlcdClassshrubScrub:functionalGroupTwigs/branches	1659	-2.029209
## nlcdClassgrasslandHerbaceous:functionalGroupWoody material	1659	-1.156293
## nlcdClassshrubScrub:functionalGroupWoody material	1659	-1.338081
##	p-value	
## (Intercept)	0.7492	
## nlcdClassgrasslandHerbaceous	0.8457	
## nlcdClassshrubScrub	0.8754	

## functionalGroupLeaves	0.8188		
## functionalGroupMixed	0.0195		
## functionalGroupNeedles	0.0000		
## functionalGroupOther	0.9297		
## functionalGroupSeeds	0.9150		
## functionalGroupTwigs/branches	0.0003		
## functionalGroupWoody material	0.0423		
## nlcdClassgrasslandHerbaceous:functionalGroupLeaves	0.8376		
## nlcdClassshrubScrub:functionalGroupLeaves	0.8170		
## nlcdClassgrasslandHerbaceous:functionalGroupMixed	0.6764		
## nlcdClassshrubScrub:functionalGroupMixed	0.4918		
## nlcdClassgrasslandHerbaceous:functionalGroupNeedles	0.0151		
## nlcdClassshrubScrub:functionalGroupNeedles	0.0001		
## nlcdClassgrasslandHerbaceous:functionalGroupOther	0.8729		
## nlcdClassshrubScrub:functionalGroupOther	0.8892		
## nlcdClassgrasslandHerbaceous:functionalGroupSeeds	0.9562		
## nlcdClassshrubScrub:functionalGroupSeeds	0.9191		
## nlcdClassgrasslandHerbaceous:functionalGroupTwigs/branches	0.2651		
## nlcdClassshrubScrub:functionalGroupTwigs/branches	0.0426		
## nlcdClassgrasslandHerbaceous:functionalGroupWoody material	0.2477		
## nlcdClassshrubScrub:functionalGroupWoody material	0.1811		
## Correlation:			
##	(Intr)	nlcdCH	nlcdCS
## nlcdClassgrasslandHerbaceous	-0.624		
## nlcdClassshrubScrub	-0.733	0.458	
## functionalGroupLeaves	-0.559	0.349	0.409
## functionalGroupMixed	-0.485	0.303	0.356
## functionalGroupNeedles	-0.579	0.361	0.424
## functionalGroupOther	-0.559	0.349	0.409
## functionalGroupSeeds	-0.559	0.349	0.409
## functionalGroupTwigs/branches	-0.571	0.356	0.418
## functionalGroupWoody material	-0.584	0.365	0.428
## nlcdClassgrasslandHerbaceous:functionalGroupLeaves	0.347	-0.586	-0.255
## nlcdClassshrubScrub:functionalGroupLeaves	0.409	-0.255	-0.569
## nlcdClassgrasslandHerbaceous:functionalGroupMixed	0.274	-0.462	-0.201
## nlcdClassshrubScrub:functionalGroupMixed	0.333	-0.208	-0.464
## nlcdClassgrasslandHerbaceous:functionalGroupNeedles	0.353	-0.595	-0.259
## nlcdClassshrubScrub:functionalGroupNeedles	0.418	-0.261	-0.582
## nlcdClassgrasslandHerbaceous:functionalGroupOther	0.342	-0.577	-0.251
## nlcdClassshrubScrub:functionalGroupOther	0.408	-0.255	-0.568
## nlcdClassgrasslandHerbaceous:functionalGroupSeeds	0.342	-0.577	-0.251
## nlcdClassshrubScrub:functionalGroupSeeds	0.407	-0.254	-0.566
## nlcdClassgrasslandHerbaceous:functionalGroupTwigs/branches	0.347	-0.586	-0.254
## nlcdClassshrubScrub:functionalGroupTwigs/branches	0.415	-0.259	-0.577
## nlcdClassgrasslandHerbaceous:functionalGroupWoody material	0.349	-0.589	-0.256
## nlcdClassshrubScrub:functionalGroupWoody material	0.420	-0.262	-0.584
##	fnctGL	fnctGM	fnctGN
## nlcdClassgrasslandHerbaceous			
## nlcdClassshrubScrub			
## functionalGroupLeaves			
## functionalGroupMixed	0.429		
## functionalGroupNeedles	0.511	0.445	
## functionalGroupOther	0.494	0.430	0.511
## functionalGroupSeeds	0.494	0.429	0.511

## functionalGroupTwigs/branches	0.504	0.439	0.522
## functionalGroupWoody material	0.516	0.449	0.535
## nlcdClassgrasslandHerbaceous:functionalGroupLeaves	-0.622	-0.267	-0.318
## nlcdClassshrubScrub:functionalGroupLeaves	-0.732	-0.314	-0.374
## nlcdClassgrasslandHerbaceous:functionalGroupMixed	-0.242	-0.564	-0.251
## nlcdClassshrubScrub:functionalGroupMixed	-0.295	-0.686	-0.305
## nlcdClassgrasslandHerbaceous:functionalGroupNeedles	-0.312	-0.272	-0.610
## nlcdClassshrubScrub:functionalGroupNeedles	-0.370	-0.322	-0.723
## nlcdClassgrasslandHerbaceous:functionalGroupOther	-0.303	-0.263	-0.313
## nlcdClassshrubScrub:functionalGroupOther	-0.361	-0.314	-0.374
## nlcdClassgrasslandHerbaceous:functionalGroupSeeds	-0.303	-0.263	-0.313
## nlcdClassshrubScrub:functionalGroupSeeds	-0.360	-0.313	-0.373
## nlcdClassgrasslandHerbaceous:functionalGroupTwigs/branches	-0.307	-0.267	-0.318
## nlcdClassshrubScrub:functionalGroupTwigs/branches	-0.367	-0.319	-0.380
## nlcdClassgrasslandHerbaceous:functionalGroupWoody material	-0.309	-0.268	-0.320
## nlcdClassshrubScrub:functionalGroupWoody material	-0.371	-0.322	-0.384
##	fnctG0	fnctGS	fncGT/
## nlcdClassgrasslandHerbaceous			
## nlcdClassshrubScrub			
## functionalGroupLeaves			
## functionalGroupMixed			
## functionalGroupNeedles			
## functionalGroupOther			
## functionalGroupSeeds	0.494		
## functionalGroupTwigs/branches	0.504	0.504	
## functionalGroupWoody material	0.516	0.517	0.528
## nlcdClassgrasslandHerbaceous:functionalGroupLeaves	-0.307	-0.307	-0.314
## nlcdClassshrubScrub:functionalGroupLeaves	-0.362	-0.362	-0.369
## nlcdClassgrasslandHerbaceous:functionalGroupMixed	-0.243	-0.242	-0.248
## nlcdClassshrubScrub:functionalGroupMixed	-0.295	-0.294	-0.301
## nlcdClassgrasslandHerbaceous:functionalGroupNeedles	-0.312	-0.312	-0.319
## nlcdClassshrubScrub:functionalGroupNeedles	-0.370	-0.370	-0.378
## nlcdClassgrasslandHerbaceous:functionalGroupOther	-0.613	-0.303	-0.309
## nlcdClassshrubScrub:functionalGroupOther	-0.731	-0.361	-0.369
## nlcdClassgrasslandHerbaceous:functionalGroupSeeds	-0.303	-0.613	-0.309
## nlcdClassshrubScrub:functionalGroupSeeds	-0.360	-0.729	-0.368
## nlcdClassgrasslandHerbaceous:functionalGroupTwigs/branches	-0.307	-0.307	-0.608
## nlcdClassshrubScrub:functionalGroupTwigs/branches	-0.367	-0.367	-0.727
## nlcdClassgrasslandHerbaceous:functionalGroupWoody material	-0.309	-0.309	-0.315
## nlcdClassshrubScrub:functionalGroupWoody material	-0.371	-0.371	-0.379
##	fncGWm	nCH:GL	nCS:GL
## nlcdClassgrasslandHerbaceous			
## nlcdClassshrubScrub			
## functionalGroupLeaves			
## functionalGroupMixed			
## functionalGroupNeedles			
## functionalGroupOther			
## functionalGroupSeeds			
## functionalGroupTwigs/branches			
## functionalGroupWoody material			
## nlcdClassgrasslandHerbaceous:functionalGroupLeaves	-0.321		
## nlcdClassshrubScrub:functionalGroupLeaves	-0.378	0.455	
## nlcdClassgrasslandHerbaceous:functionalGroupMixed	-0.253	0.406	0.178
## nlcdClassshrubScrub:functionalGroupMixed	-0.308	0.183	0.410

## nlcdClassgrasslandHerbaceous:functionalGroupNeedles	-0.326	0.524	0.229
## nlcdClassshrubScrub:functionalGroupNeedles	-0.387	0.230	0.514
## nlcdClassgrasslandHerbaceous:functionalGroupOther	-0.316	0.508	0.222
## nlcdClassshrubScrub:functionalGroupOther	-0.377	0.224	0.502
## nlcdClassgrasslandHerbaceous:functionalGroupSeeds	-0.317	0.508	0.222
## nlcdClassshrubScrub:functionalGroupSeeds	-0.376	0.224	0.500
## nlcdClassgrasslandHerbaceous:functionalGroupTwigs/branches	-0.321	0.515	0.225
## nlcdClassshrubScrub:functionalGroupTwigs/branches	-0.384	0.228	0.510
## nlcdClassgrasslandHerbaceous:functionalGroupWoody material	-0.597	0.518	0.226
## nlcdClassshrubScrub:functionalGroupWoody material	-0.719	0.231	0.516
##	nCH:GM	nCS:GM	nCH:GN
## nlcdClassgrasslandHerbaceous			
## nlcdClassshrubScrub			
## functionalGroupLeaves			
## functionalGroupMixed			
## functionalGroupNeedles			
## functionalGroupOther			
## functionalGroupSeeds			
## functionalGroupTwigs/branches			
## functionalGroupWoody material			
## nlcdClassgrasslandHerbaceous:functionalGroupLeaves			
## nlcdClassshrubScrub:functionalGroupLeaves			
## nlcdClassgrasslandHerbaceous:functionalGroupMixed			
## nlcdClassshrubScrub:functionalGroupMixed	0.387		
## nlcdClassgrasslandHerbaceous:functionalGroupNeedles	0.414	0.186	
## nlcdClassshrubScrub:functionalGroupNeedles	0.182	0.419	0.441
## nlcdClassgrasslandHerbaceous:functionalGroupOther	0.401	0.181	0.517
## nlcdClassshrubScrub:functionalGroupOther	0.177	0.409	0.228
## nlcdClassgrasslandHerbaceous:functionalGroupSeeds	0.402	0.180	0.517
## nlcdClassshrubScrub:functionalGroupSeeds	0.177	0.408	0.227
## nlcdClassgrasslandHerbaceous:functionalGroupTwigs/branches	0.407	0.183	0.524
## nlcdClassshrubScrub:functionalGroupTwigs/branches	0.180	0.416	0.232
## nlcdClassgrasslandHerbaceous:functionalGroupWoody material	0.409	0.184	0.527
## nlcdClassshrubScrub:functionalGroupWoody material	0.182	0.420	0.235
##	nCS:GN	nCH:GO	nCS:GO
## nlcdClassgrasslandHerbaceous			
## nlcdClassshrubScrub			
## functionalGroupLeaves			
## functionalGroupMixed			
## functionalGroupNeedles			
## functionalGroupOther			
## functionalGroupSeeds			
## functionalGroupTwigs/branches			
## functionalGroupWoody material			
## nlcdClassgrasslandHerbaceous:functionalGroupLeaves			
## nlcdClassshrubScrub:functionalGroupLeaves			
## nlcdClassgrasslandHerbaceous:functionalGroupMixed			
## nlcdClassshrubScrub:functionalGroupMixed			
## nlcdClassgrasslandHerbaceous:functionalGroupNeedles			
## nlcdClassshrubScrub:functionalGroupNeedles			
## nlcdClassgrasslandHerbaceous:functionalGroupOther	0.227		
## nlcdClassshrubScrub:functionalGroupOther	0.513	0.448	
## nlcdClassgrasslandHerbaceous:functionalGroupSeeds	0.227	0.501	0.221
## nlcdClassshrubScrub:functionalGroupSeeds	0.512	0.221	0.499

```

## nlcdClassgrasslandHerbaceous:functionalGroupTwigs/branches 0.230 0.508 0.224
## nlcdClassshrubScrub:functionalGroupTwigs/branches 0.521 0.225 0.509
## nlcdClassgrasslandHerbaceous:functionalGroupWoody material 0.231 0.511 0.225
## nlcdClassshrubScrub:functionalGroupWoody material 0.528 0.227 0.515
## nCH:GS nCS:GS nCH:GT
## nlcdClassgrasslandHerbaceous
## nlcdClassshrubScrub
## functionalGroupLeaves
## functionalGroupMixed
## functionalGroupNeedles
## functionalGroupOther
## functionalGroupSeeds
## functionalGroupTwigs/branches
## functionalGroupWoody material
## nlcdClassgrasslandHerbaceous:functionalGroupLeaves
## nlcdClassshrubScrub:functionalGroupLeaves
## nlcdClassgrasslandHerbaceous:functionalGroupMixed
## nlcdClassshrubScrub:functionalGroupMixed
## nlcdClassgrasslandHerbaceous:functionalGroupNeedles
## nlcdClassshrubScrub:functionalGroupNeedles
## nlcdClassgrasslandHerbaceous:functionalGroupOther
## nlcdClassshrubScrub:functionalGroupOther
## nlcdClassgrasslandHerbaceous:functionalGroupSeeds
## nlcdClassshrubScrub:functionalGroupSeeds 0.447
## nlcdClassgrasslandHerbaceous:functionalGroupTwigs/branches 0.508 0.224
## nlcdClassshrubScrub:functionalGroupTwigs/branches 0.225 0.507 0.442
## nlcdClassgrasslandHerbaceous:functionalGroupWoody material 0.511 0.225 0.518
## nlcdClassshrubScrub:functionalGroupWoody material 0.228 0.514 0.231
## nCS:GT nCH:Gm
## nlcdClassgrasslandHerbaceous
## nlcdClassshrubScrub
## functionalGroupLeaves
## functionalGroupMixed
## functionalGroupNeedles
## functionalGroupOther
## functionalGroupSeeds
## functionalGroupTwigs/branches
## functionalGroupWoody material
## nlcdClassgrasslandHerbaceous:functionalGroupLeaves
## nlcdClassshrubScrub:functionalGroupLeaves
## nlcdClassgrasslandHerbaceous:functionalGroupMixed
## nlcdClassshrubScrub:functionalGroupMixed
## nlcdClassgrasslandHerbaceous:functionalGroupNeedles
## nlcdClassshrubScrub:functionalGroupNeedles
## nlcdClassgrasslandHerbaceous:functionalGroupOther
## nlcdClassshrubScrub:functionalGroupOther
## nlcdClassgrasslandHerbaceous:functionalGroupSeeds
## nlcdClassshrubScrub:functionalGroupSeeds
## nlcdClassgrasslandHerbaceous:functionalGroupTwigs/branches
## nlcdClassshrubScrub:functionalGroupTwigs/branches
## nlcdClassgrasslandHerbaceous:functionalGroupWoody material 0.229
## nlcdClassshrubScrub:functionalGroupWoody material 0.523 0.429
##
## Standardized Within-Group Residuals:

```

```
##           Min           Q1           Med           Q3           Max
## -1.96496855 -0.23842984 -0.01535880  0.09027291 14.27434811
##
## Number of Observations: 1692
## Number of Groups: 12
```

```
rsquared(litter.mixed) # which one is the rsquared?? conditional or marginal??
```

```
## Response family link method Marginal Conditional
## 1 dryMass gaussian identity none 0.2465822 0.2679023
```

b. continued... How much more variance is explained by adding the random effect to the model?

Answer: about 2 % more variance is explained by including the random effect (difference between the marginal and conditional r squared)

c. Run the same model without the random effect.

d. Run an anova on the two tests.

9c.

```
litter.fixed <- lm(data = litter.data, dryMass ~ nlcdClass * functionalGroup)
summary(litter.fixed)
```

```
##
## Call:
## lm(formula = dryMass ~ nlcdClass * functionalGroup, data = litter.data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -6.612  -0.480  -0.058  -0.005  49.051
##
## Coefficients:
##                                     Estimate Std. Error
## (Intercept)                        0.11963    0.39070
## nlcdClassgrasslandHerbaceous       -0.11420    0.64223
## nlcdClassshrubScrub                -0.10412    0.53838
## functionalGroupLeaves              -0.10360    0.55606
## functionalGroupMixed                1.50475    0.63800
## functionalGroupNeedles              7.31226    0.53696
## functionalGroupOther               -0.03482    0.55606
## functionalGroupSeeds               -0.04616    0.55606
## functionalGroupTwigs/branches       1.95967    0.54434
## functionalGroupWoody material       1.08431    0.53156
## nlcdClassgrasslandHerbaceous:functionalGroupLeaves  0.12865    0.89410
## nlcdClassshrubScrub:functionalGroupLeaves  0.14703    0.75915
## nlcdClassgrasslandHerbaceous:functionalGroupMixed -0.38118    1.13024
## nlcdClassshrubScrub:functionalGroupMixed  0.74593    0.93038
## nlcdClassgrasslandHerbaceous:functionalGroupNeedles -2.13880    0.87993
## nlcdClassshrubScrub:functionalGroupNeedles -2.92148    0.74258
## nlcdClassgrasslandHerbaceous:functionalGroupOther  0.12606    0.90743
## nlcdClassshrubScrub:functionalGroupOther  0.08589    0.76101
## nlcdClassgrasslandHerbaceous:functionalGroupSeeds  0.04615    0.90743
## nlcdClassshrubScrub:functionalGroupSeeds  0.05944    0.76295
## nlcdClassgrasslandHerbaceous:functionalGroupTwigs/branches -1.01519    0.89462
## nlcdClassshrubScrub:functionalGroupTwigs/branches -1.49559    0.74881
## nlcdClassgrasslandHerbaceous:functionalGroupWoody material -1.04086    0.88971
## nlcdClassshrubScrub:functionalGroupWoody material -0.97185    0.73957
```

```
## t value Pr(>|t|)
## (Intercept) 0.306 0.759502
## nlcdClassgrasslandHerbaceous -0.178 0.858888
## nlcdClassshrubScrub -0.193 0.846673
## functionalGroupLeaves -0.186 0.852224
## functionalGroupMixed 2.359 0.018462 *
## functionalGroupNeedles 13.618 < 2e-16 ***
## functionalGroupOther -0.063 0.950081
## functionalGroupSeeds -0.083 0.933846
## functionalGroupTwigs/branches 3.600 0.000327 ***
## functionalGroupWoody material 2.040 0.041519 *
## nlcdClassgrasslandHerbaceous:functionalGroupLeaves 0.144 0.885611
## nlcdClassshrubScrub:functionalGroupLeaves 0.194 0.846453
## nlcdClassgrasslandHerbaceous:functionalGroupMixed -0.337 0.735969
## nlcdClassshrubScrub:functionalGroupMixed 0.802 0.422814
## nlcdClassgrasslandHerbaceous:functionalGroupNeedles -2.431 0.015177 *
## nlcdClassshrubScrub:functionalGroupNeedles -3.934 8.69e-05 ***
## nlcdClassgrasslandHerbaceous:functionalGroupOther 0.139 0.889531
## nlcdClassshrubScrub:functionalGroupOther 0.113 0.910155
## nlcdClassgrasslandHerbaceous:functionalGroupSeeds 0.051 0.959441
## nlcdClassshrubScrub:functionalGroupSeeds 0.078 0.937915
## nlcdClassgrasslandHerbaceous:functionalGroupTwigs/branches -1.135 0.256634
## nlcdClassshrubScrub:functionalGroupTwigs/branches -1.997 0.045956 *
## nlcdClassgrasslandHerbaceous:functionalGroupWoody material -1.170 0.242213
## nlcdClassshrubScrub:functionalGroupWoody material -1.314 0.189001
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.494 on 1668 degrees of freedom
## Multiple R-squared: 0.2516, Adjusted R-squared: 0.2413
## F-statistic: 24.38 on 23 and 1668 DF, p-value: < 2.2e-16
```

```
anova(litter.mixed, litter.fixed)
```

```
## Model df AIC BIC logLik Test L.Ratio p-value
## litter.mixed 1 26 9038.575 9179.479 -4493.287
## litter.fixed 2 25 9058.088 9193.573 -4504.044 1 vs 2 21.51338 <.0001
```

d. continued... Is the mixed effects model a better model than the fixed effects model? How do you know?

Answer: the mixed model is better because the AIC score is lower