## Solution to challenge problem 2: GLS vs. PIC regression

We start by loading our packages:

```
library(ape)
library(nlme)
library(geiger)
```

Then our data from file:

```
datos<-read.csv("Barbetdata.csv",header=TRUE,row.names=1)
arbol<-read.nexus("BarbetTree.nex")</pre>
```

Run name check to verify that we have the same data in the tree and data set:

```
obj<-name.check(arbol,datos)
arbol.cortado<-drop.tip(arbol, obj$tree_not_data)
name.check(arbol.cortado, datos)
```

```
## [1] "OK"
```

Now let's make a correlation structure for our PGLS model and fit it:

```
bm<-corBrownian(1,arbol.cortado)
pgls.model<-gls(wing~Lnalt,data=datos,correlation=bm)
summary(pgls.model)</pre>
```

```
## Generalized least squares fit by REML
##
     Model: wing ~ Lnalt
##
     Data: datos
##
           AIC
                     BIC
                           logLik
##
     -41.52099 -37.21902 23.76049
##
## Correlation Structure: corBrownian
##
    Formula: ~1
##
   Parameter estimate(s):
## numeric(0)
##
## Coefficients:
##
                   Value Std.Error
                                      t-value p-value
## (Intercept) 4.808792 0.23025431 20.884700 0.0000
## Lnalt
               -0.053410 0.03332457 -1.602729
                                               0.1191
##
##
   Correlation:
##
         (Intr)
## Lnalt -0.879
##
## Standardized residuals:
##
          Min
                      Q1
                                Med
                                             Q3
                                                       Max
## -0.8708169
              0.3897004 0.6685024 0.9361152
                                                 2.7445358
##
## Residual standard error: 0.1986708
## Degrees of freedom: 33 total; 31 residual
```

Next, we can fit our PIC regression:

```
wing<-setNames(datos$wing,rownames(datos))
Lnalt<-setNames(datos$Lnalt,rownames(datos))
pic.wing<-pic(wing,arbol.cortado)
pic.Lnalt<-pic(Lnalt,arbol.cortado)
pic.model<-lm(pic.wing~pic.Lnalt+0)
summary(pic.model)</pre>
```

```
##
## Call:
## lm(formula = pic.wing ~ pic.Lnalt + 0)
##
## Residuals:
##
         Min
                    1Q
                          Median
                                        3Q
                                                 Max
## -0.061160 -0.018176 -0.001805
                                 0.025447 0.086057
##
## Coefficients:
##
             Estimate Std. Error t value Pr(>|t|)
## pic.Lnalt -0.05341 0.03332 -1.603
                                            0.119
##
## Residual standard error: 0.03931 on 31 degrees of freedom
## Multiple R-squared: 0.07652,
                                  Adjusted R-squared: 0.04673
## F-statistic: 2.569 on 1 and 31 DF, p-value: 0.1191
```

What have we found?

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