

Lab 2: Independant Contrasts

Comparative Biology and Macroevolution

April 12, 2019

Classwork

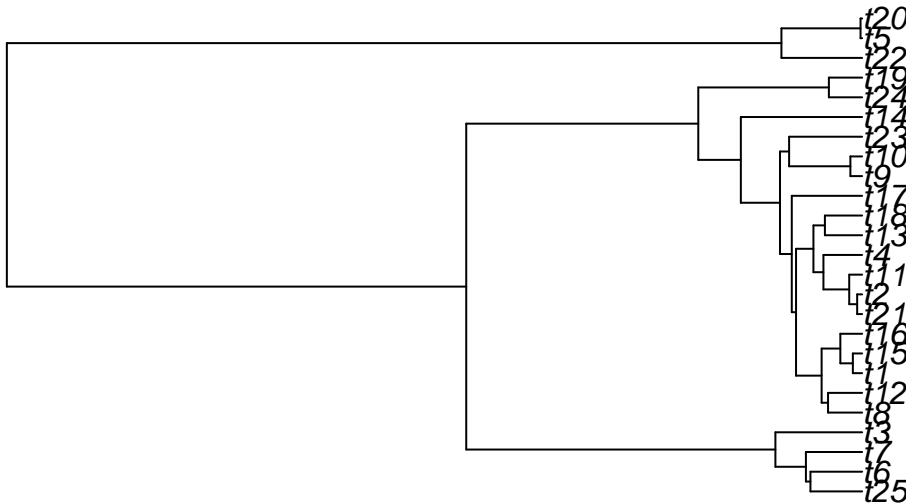
One kind of analysis that is frequently done by evolutionary biologist is trait correlations. For example, we might ask whether brain size and body size are correlated, or if the cribiform plate surface area correlates to the number of olfactory receptor genes. While the analysis you may be more familiar with deals with comparing individuals from the same species, in this class, we will be thinking of trait correlations between different species.

One problem with a naive analysis of trait data between species is that species are not independent. In his 1985 paper, Felsenstein points out that species are part of a “hierarchically structured phylogeny” and therefore not independent. Since this insight, there have been multiple approaches that have been developed to help mitigate this problem. In this lab, we will be using the first *fully* phylogenetic method, termed by Felsenstein as independent contrasts (IC), to incorporate a phylogenetic perspective on comparative data.

Import Data

```
library(ape)
```

```
tree <- read.tree("lab2_tree_new.tre")  
plot(tree)
```



```
traits <- read.csv("lab2_data1.csv", stringsAsFactors = F)  
traits
```

##	Taxon	trait1	trait2
## 1	t1	18.32059	40.62082
## 2	t10	17.58344	41.70659
## 3	t11	18.45661	40.78300
## 4	t12	17.89541	41.59028
## 5	t13	17.94329	41.81583
## 6	t14	17.95515	40.93829

```
## 7      t15 18.49101 40.78501
## 8      t16 18.50241 40.74782
## 9      t17 17.57091 41.08303
## 10     t18 18.25691 41.74148
## 11     t19 18.94612 40.05859
## 12      t2 18.71446 40.87573
## 13     t20 16.76495 37.15394
## 14     t21 18.90497 40.41419
## 15     t22 17.34779 37.23577
## 16     t23 17.94877 41.79086
## 17     t24 18.50319 40.05271
## 18     t25 17.81249 39.34886
## 19      t3 18.49510 39.82282
## 20      t4 18.23416 41.28766
## 21      t5 16.85516 36.93270
## 22      t6 18.14229 40.01092
## 23      t7 17.68522 39.74654
## 24      t8 18.54884 40.72576
## 25      t9 18.10201 41.73898
```

Check if the `tip.label` in the `phylo` object have the same row order as `traits$Taxon`

```
cbind(traits$Taxon, tree$tip.label)
```

```
##      [,1] [,2]
## [1,] "t1"  "t25"
## [2,] "t10" "t6"
## [3,] "t11" "t7"
## [4,] "t12" "t3"
## [5,] "t13" "t8"
## [6,] "t14" "t12"
## [7,] "t15" "t1"
## [8,] "t16" "t15"
## [9,] "t17" "t16"
## [10,] "t18" "t21"
## [11,] "t19" "t2"
## [12,] "t2"  "t11"
## [13,] "t20" "t4"
## [14,] "t21" "t13"
## [15,] "t22" "t18"
## [16,] "t23" "t17"
## [17,] "t24" "t9"
## [18,] "t25" "t10"
## [19,] "t3"  "t23"
## [20,] "t4"  "t14"
## [21,] "t5"  "t24"
## [22,] "t6"  "t19"
## [23,] "t7"  "t22"
## [24,] "t8"  "t5"
## [25,] "t9"  "t20"
```

The species data in the `traits` data.frame must be in the same row order as the names of the species in the tree object (species names are in `tree$tip.label`). To accomplish this, begin by changing the `rownames()` in the `traits` data.frame to the tip labels

```
rownames(traits) <- traits$Taxon
```

Run the following command to ensure that everything matches

```
traits <- traits[match(tree$tip.label, rownames(traits)), ]
traits
```

```
##      Taxon  trait1  trait2
## t25    t25 17.81249 39.34886
## t6      t6 18.14229 40.01092
## t7      t7 17.68522 39.74654
## t3      t3 18.49510 39.82282
## t8      t8 18.54884 40.72576
## t12     t12 17.89541 41.59028
## t1      t1 18.32059 40.62082
## t15     t15 18.49101 40.78501
## t16     t16 18.50241 40.74782
## t21     t21 18.90497 40.41419
## t2      t2 18.71446 40.87573
## t11     t11 18.45661 40.78300
## t4      t4 18.23416 41.28766
## t13     t13 17.94329 41.81583
## t18     t18 18.25691 41.74148
## t17     t17 17.57091 41.08303
## t9      t9 18.10201 41.73898
## t10     t10 17.58344 41.70659
## t23     t23 17.94877 41.79086
## t14     t14 17.95515 40.93829
## t24     t24 18.50319 40.05271
## t19     t19 18.94612 40.05859
## t22     t22 17.34779 37.23577
## t5      t5 16.85516 36.93270
## t20     t20 16.76495 37.15394
```

Check again to make sure that `tip.label` in the tree phylo object have same row order as `traits$Taxon`

```
cbind(traits$Taxon, tree$tip.label)
```

```
##      [,1] [,2]
## [1,] "t25" "t25"
## [2,] "t6"  "t6"
## [3,] "t7"  "t7"
## [4,] "t3"  "t3"
## [5,] "t8"  "t8"
## [6,] "t12" "t12"
## [7,] "t1"  "t1"
## [8,] "t15" "t15"
## [9,] "t16" "t16"
## [10,] "t21" "t21"
## [11,] "t2"  "t2"
## [12,] "t11" "t11"
## [13,] "t4"  "t4"
## [14,] "t13" "t13"
## [15,] "t18" "t18"
## [16,] "t17" "t17"
## [17,] "t9"  "t9"
```

```
## [18,] "t10" "t10"
## [19,] "t23" "t23"
## [20,] "t14" "t14"
## [21,] "t24" "t24"
## [22,] "t19" "t19"
## [23,] "t22" "t22"
## [24,] "t5"  "t5"
## [25,] "t20" "t20"
```

Okay now we're ready for analysis!

Naive analysis of trait data

Extract trait data from `traits` data.frame

```
trait1 <- traits$trait1
trait1
```

```
## [1] 17.81249 18.14229 17.68522 18.49510 18.54884 17.89541 18.32059
## [8] 18.49101 18.50241 18.90497 18.71446 18.45661 18.23416 17.94329
## [15] 18.25691 17.57091 18.10201 17.58344 17.94877 17.95515 18.50319
## [22] 18.94612 17.34779 16.85516 16.76495
```

```
trait2 <- traits$trait2
trait2
```

```
## [1] 39.34886 40.01092 39.74654 39.82282 40.72576 41.59028 40.62082
## [8] 40.78501 40.74782 40.41419 40.87573 40.78300 41.28766 41.81583
## [15] 41.74148 41.08303 41.73898 41.70659 41.79086 40.93829 40.05271
## [22] 40.05859 37.23577 36.93270 37.15394
```

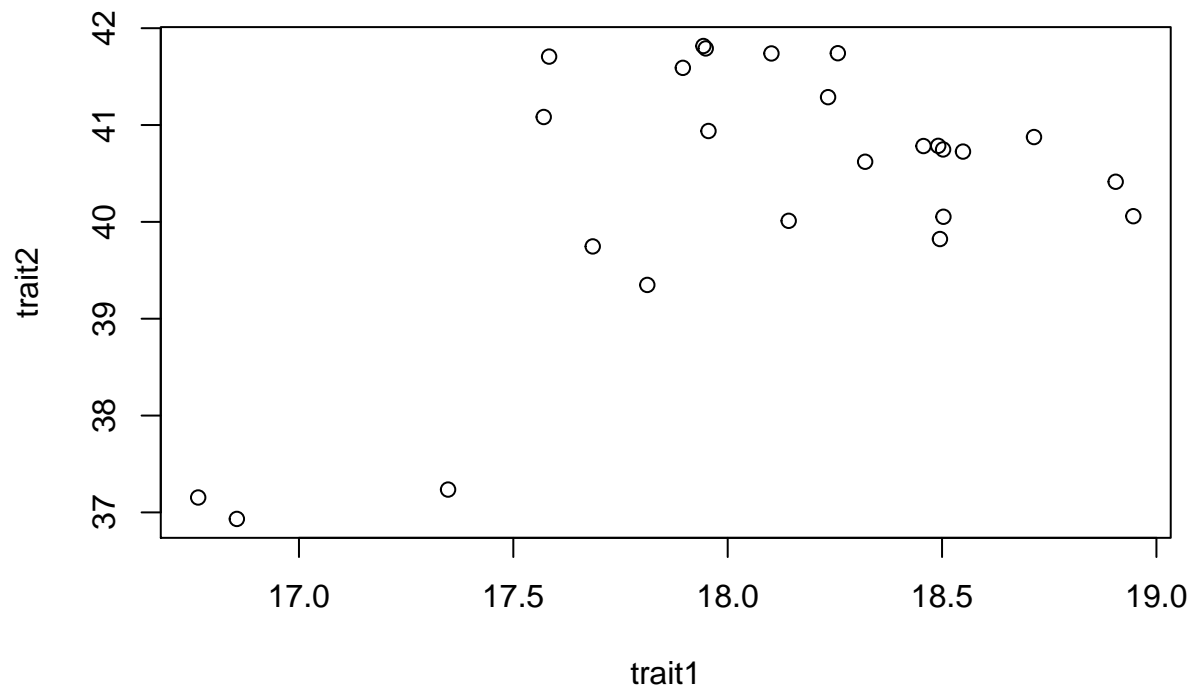
FYI: sometimes we need to compute relative traits (e.g., brain size adjusted for body size). This is easily done by dividing columns

```
relative_trait1 <- trait1/trait2
relative_trait1
```

```
## [1] 0.4526812 0.4534335 0.4449499 0.4644348 0.4554573 0.4302788 0.4510148
## [8] 0.4533777 0.4540711 0.4677805 0.4578379 0.4525565 0.4416371 0.4291027
## [15] 0.4373805 0.4276926 0.4336956 0.4215987 0.4294902 0.4385907 0.4619711
## [22] 0.4729603 0.4658906 0.4563748 0.4512293
```

Plot the trait data to visualize the data

```
plot(trait1, trait2)
```



Transform the trait data with natural log. Plot the $\ln(\text{data})$

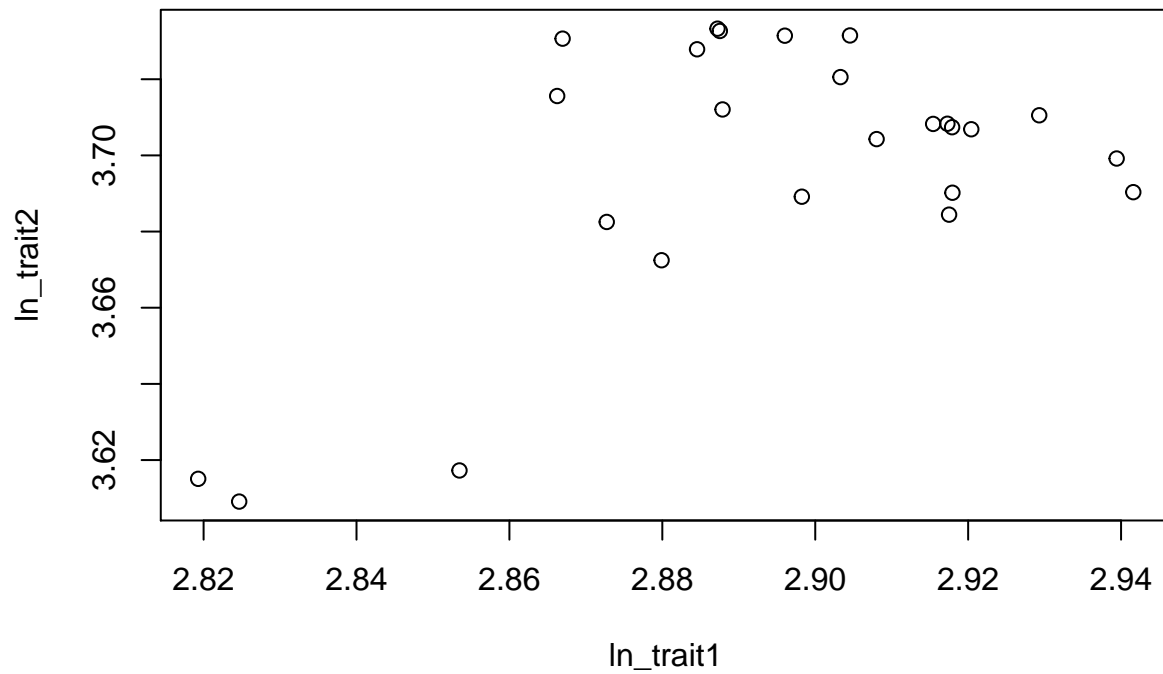
```
ln_trait1 <- log(trait1)
ln_trait1
```

```
## [1] 2.879900 2.898246 2.872729 2.917506 2.920407 2.884544 2.908026
## [8] 2.917285 2.917901 2.939425 2.929296 2.915423 2.903297 2.887216
## [15] 2.904544 2.866245 2.896023 2.866958 2.887521 2.887877 2.917943
## [22] 2.941599 2.853465 2.824657 2.819290
```

```
ln_trait2 <- log(trait2)
ln_trait2
```

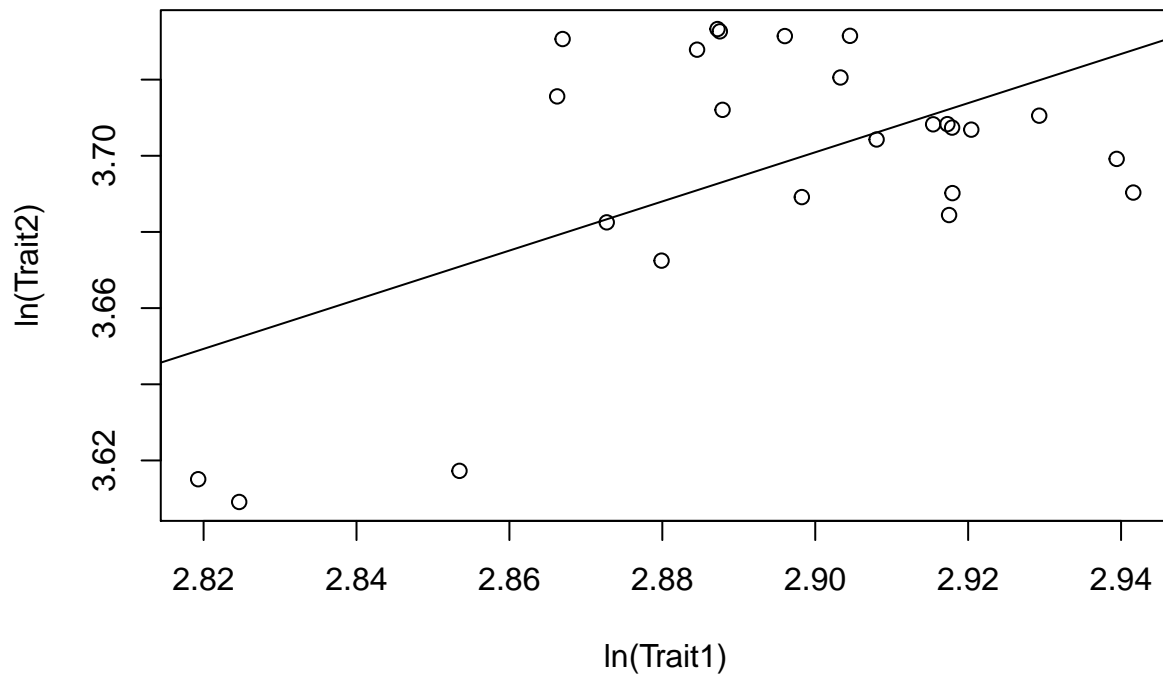
```
## [1] 3.672467 3.689153 3.682523 3.684440 3.706861 3.727866 3.704281
## [8] 3.708315 3.707402 3.699181 3.710536 3.708265 3.720564 3.733275
## [15] 3.731495 3.715595 3.731436 3.730659 3.732678 3.712066 3.690196
## [22] 3.690343 3.617270 3.609097 3.615070
```

```
plot(ln_trait1, ln_trait2)
```



Fit a linear model to the $\ln(\text{data})$

```
fit <- lm(log(trait2) ~ log(trait1))
plot(ln_trait1, ln_trait2, xlab = "ln(Trait1)", ylab = "ln(Trait2)")
abline(fit)
```



Question: What is the relationship between trait 1 and trait 2?

```
summary(fit)
```

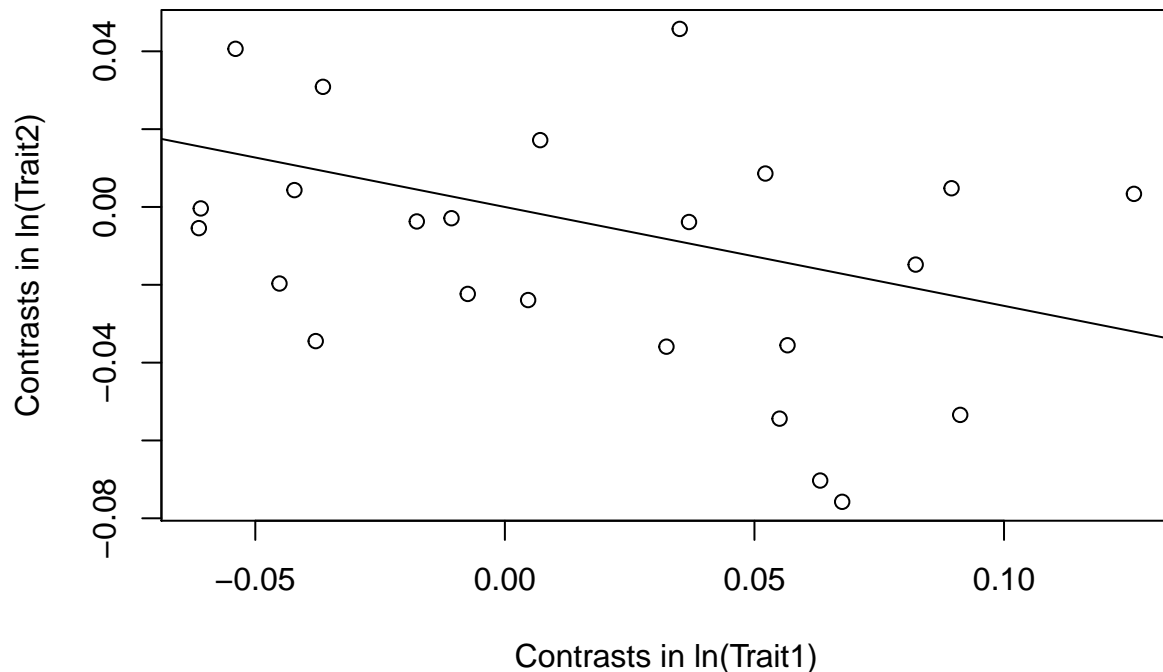
```
##
## Call:
```

```
## lm(formula = log(trait2) ~ log(trait1))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.053624 -0.022307 -0.003764  0.027639  0.051058
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   1.8295     0.5656   3.234  0.00366 **
## log(trait1)    0.6453     0.1954   3.302  0.00311 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.03014 on 23 degrees of freedom
## Multiple R-squared:  0.3217, Adjusted R-squared:  0.2922
## F-statistic: 10.91 on 1 and 23 DF,  p-value: 0.003112
```

Question: What can we conclude from the naive analysis about the relationship between trait 1 and trait 2?

Analysis with PICs

```
PIC_ln_trait1 <- pic(ln_trait1, tree)
PIC_ln_trait2 <- pic(ln_trait2, tree)
PIC_fit <- lm(PIC_ln_trait2 ~ PIC_ln_trait1 - 1) # "-1" removes intercept term
plot(PIC_ln_trait1, PIC_ln_trait2,
     xlab = "Contrasts in ln(Trait1)",
     ylab = "Contrasts in ln(Trait2)")
abline(PIC_fit)
```



Question: What can we conclude about the relationship between the PICs for trait 1 and trait 2?

```
summary(PIC_fit)
```

```
##
```

```
## Call:
## lm(formula = PIC_ln_trait2 ~ PIC_ln_trait1 - 1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.05859 -0.02832 -0.01203  0.01963  0.05466
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## PIC_ln_trait1  -0.2541      0.1112  -2.285   0.0318 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.03099 on 23 degrees of freedom
## Multiple R-squared:  0.1851, Adjusted R-squared:  0.1496
## F-statistic: 5.223 on 1 and 23 DF,  p-value: 0.03184
```

Question: Why do we fit the linear model without an intercept term?

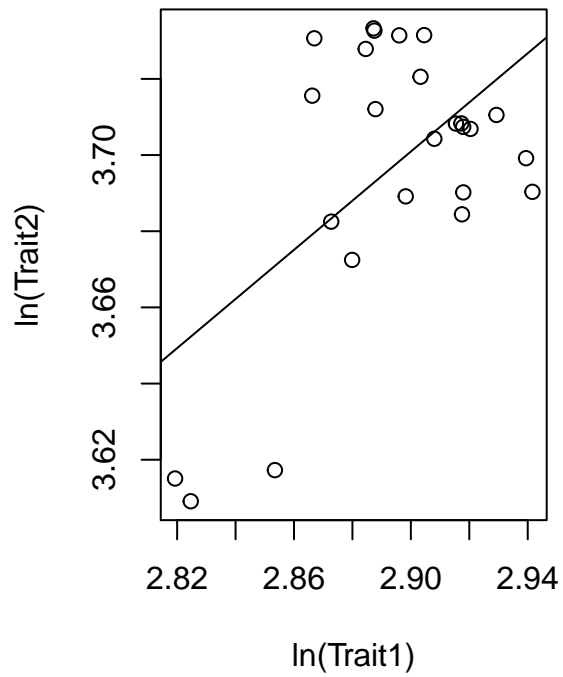
Prepare plots for a report

```
par(mfrow = c(1, 2)) #This command allows you to plot two figures side-by-side.

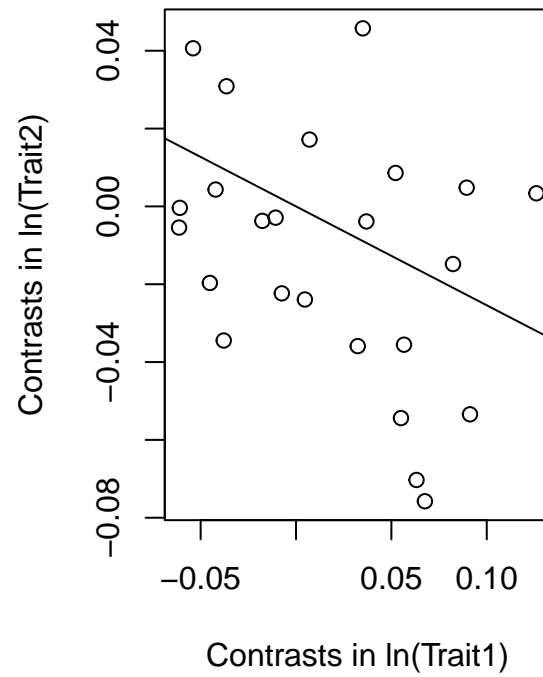
# Plotting naive analysis
plot(ln_trait1, ln_trait2,
     xlab = "ln(Trait1)", ylab = "ln(Trait2)",
     main = "Naive analysis")
abline(fit)

# Plotting Analysis using PICs
plot(PIC_ln_trait1, PIC_ln_trait2,
     xlab = "Contrasts in ln(Trait1)", ylab = "Contrasts in ln(Trait2)",
     main = "Analysis using PICs")
abline(PIC_fit)
```

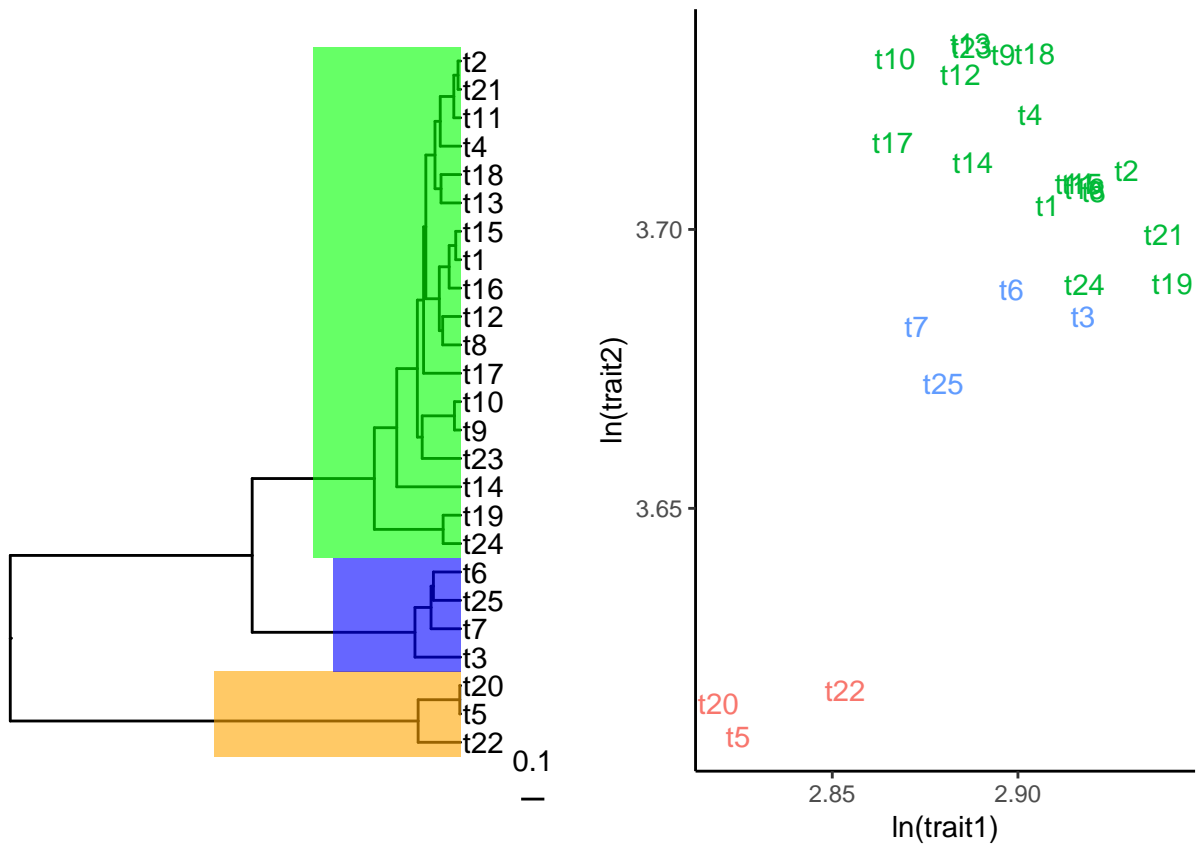

Naive analysis



Analysis using PICs



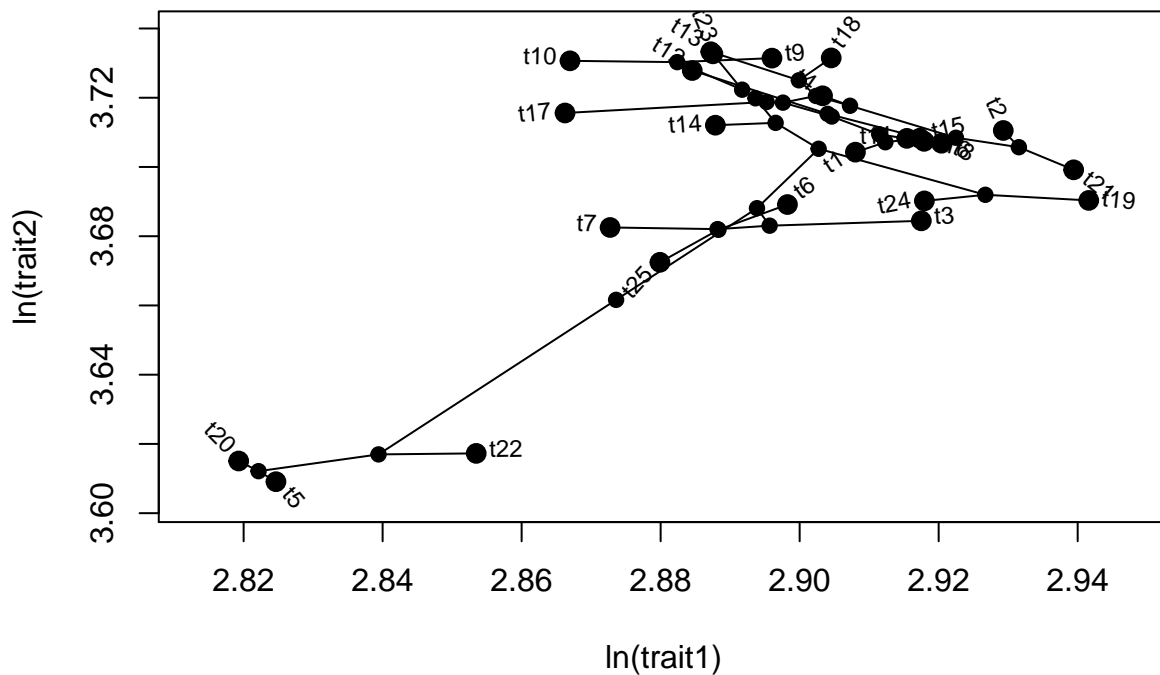
Understanding the intuition



Some Fancy stuff you can do

```
library(phytools)

## Loading required package: maps
##
## Attaching package: 'phytools'
## The following objects are masked from 'package:ggtree':
##
##   read.newick, reroot
ln_traits = cbind(ln_trait1, ln_trait2)
rownames(ln_traits) <- traits$Taxon
phylogenospace(tree = tree, ln_traits, xlab = "ln(trait1)", ylab = "ln(trait2)")
```



Homework

Courtesy of Tyler McCraney

Why do some bats have larger testes than others?

Reanalyze the trait data *batTraits.csv* from Hosken (1998) using the updated bat tree *batTree* by Shi and Rabosky (2015).

Use relative testes mass for your analysis (*i.e.*, testes mass / body mass)

Conduct both a (1) naive analysis, and (2) proper analyses using PICs

Report the p-value and adjusted R-squared of each analysis in the text of your lab report