

# Phylogenetically Independent Contrasts

## PACKAGES

```
library(ape)
library(phytools)
```

```
## Loading required package: maps
```

## CHALLENGE PROBLEM 1

Fit a OLS (non-phylogenetic) regression model for  $y \sim x$  and a phylogenetic regression model using contrasts. What do you find? Why?

OLS regression model:

```
obj<-read.csv("pic-exercise-data.csv",row.names=1); head(obj)
```

```
##           x           y
## t64  0.8420183 -1.375677
## t4   -0.7992242 -3.132304
## t14 -0.7434011 -3.211733
## t18 -0.9669659 -3.737145
## t92 -0.7723297 -3.386689
## t35 -0.1333343 -2.110647
```

```
fit.ols<-lm(y~x,data=obj); fit.ols
```

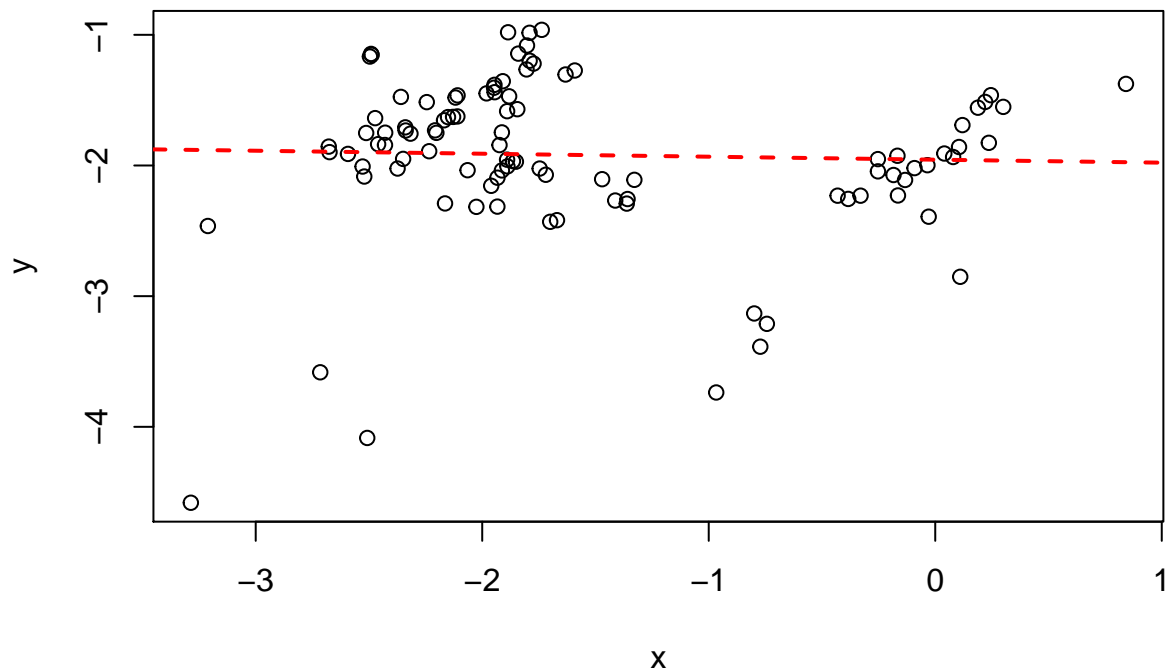
```
##
## Call:
## lm(formula = y ~ x, data = obj)
##
## Coefficients:
## (Intercept)          x
##   -1.95546      -0.02282
```

```
summary(fit.ols)
```

```
##
## Call:
## lm(formula = y ~ x, data = obj)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.70042 -0.18339  0.04348  0.41484  0.95401
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -1.95546    0.12017 -16.272  <2e-16 ***
## x           -0.02282    0.06582  -0.347    0.73
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.6289 on 98 degrees of freedom
```

```
## Multiple R-squared:  0.001225,   Adjusted R-squared:  -0.008967
## F-statistic: 0.1202 on 1 and 98 DF,  p-value: 0.7296
```

```
plot(obj[,c("x","y")])
abline(fit.ols,lwd=2,lty="dashed",col="red")
```



Phylogenetic regression model using contrasts:

```
ex.tree<-read.tree("pic-exercise-tree.tre")
x.data<-setNames(obj[, "x"],rownames(obj))
y.data<-setNames(obj[, "y"],rownames(obj))
pic.x<-pic(x.data,ex.tree) #categorize buccal length by phylogeny
pic.y<-pic(y.data,ex.tree) #categorize gape width by phylogeny
fit.pic<-lm(pic.y~pic.x+0) #fit linear model with phylogeny considered
fit.pic
```

```
##
## Call:
## lm(formula = pic.y ~ pic.x + 0)
##
## Coefficients:
##  pic.x
## 0.7039
```

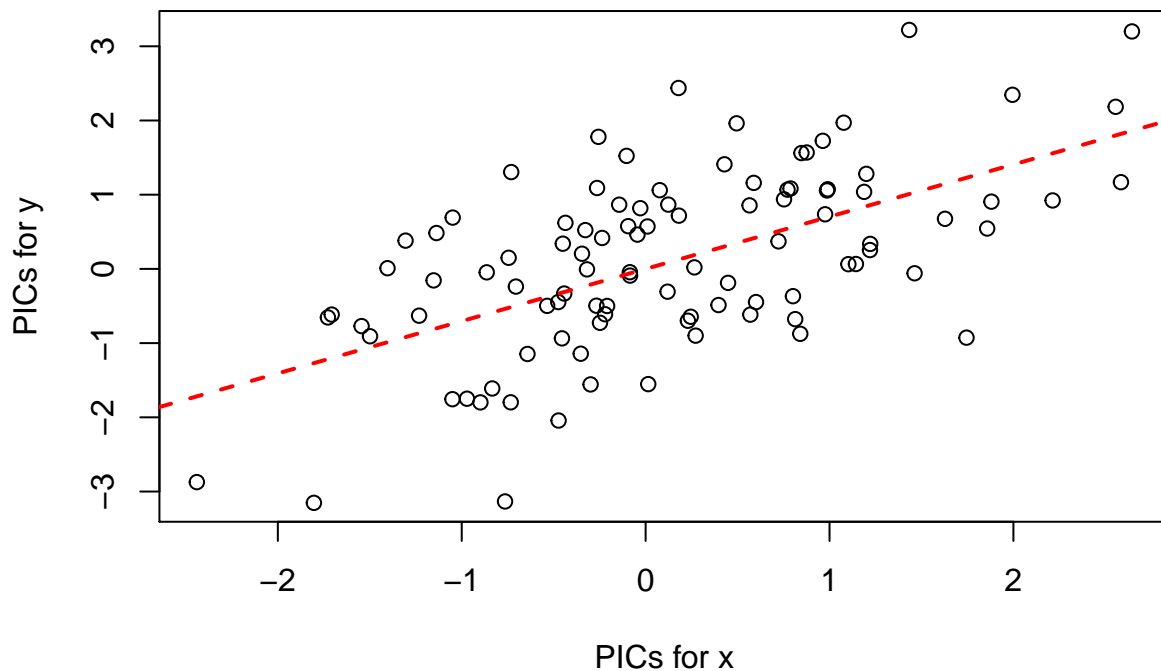
```
summary(fit.pic)
```

```
##
## Call:
## lm(formula = pic.y ~ pic.x + 0)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.5957 -0.7257  0.2020  0.7100  2.3120
##
## Coefficients:
```

```
##           Estimate Std. Error t value Pr(>|t|)
## pic.x  0.70387    0.09648   7.296 7.81e-11 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9944 on 98 degrees of freedom
## Multiple R-squared:  0.352, Adjusted R-squared:  0.3454
## F-statistic: 53.23 on 1 and 98 DF,  p-value: 7.808e-11
```

Plot linear model over scatterplot with phylogeny considered:

```
plot(pic.x,pic.y,xlab="PICs for x",ylab="PICs for y")
abline(fit.pic,lwd=2,lty="dashed",col="red")
```



## TUTORIAL

### Linear Model Basics

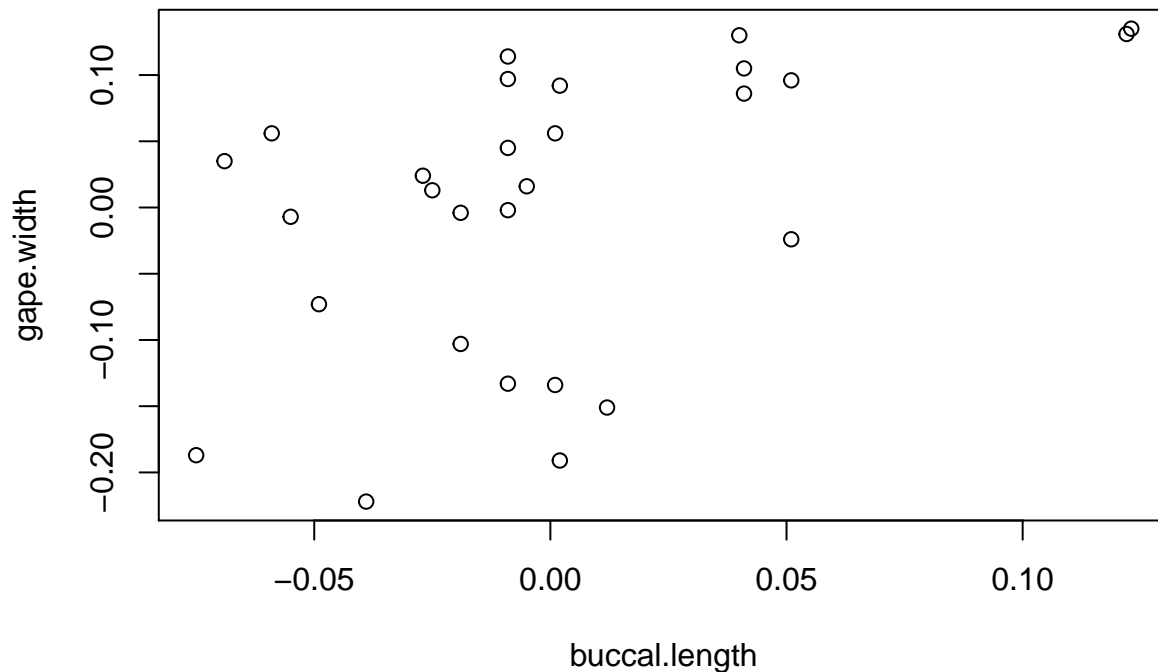
Fitting linear regressions to first gape width and buccal length. View Data:

```
obj<-read.csv("Centrarchidae.csv",row.names=1); head(obj)
```

```
##           feeding.mode gape.width buccal.length
## Acantharchus_pomotis    pisc     0.114      -0.009
## Lepomis_gibbosus        non    -0.133      -0.009
## Lepomis_microlophus     non    -0.151       0.012
## Lepomis_punctatus       non    -0.103     -0.019
## Lepomis_miniatus        non    -0.134       0.001
## Lepomis_auritus         non    -0.222     -0.039
```

Plot effect of buccal length on gape width

```
plot(obj[,c("buccal.length", "gape.width")])
```



Fit OLS regression model:

```
fit.ols<-lm(gape.width~buccal.length,data=obj); fit.ols
```

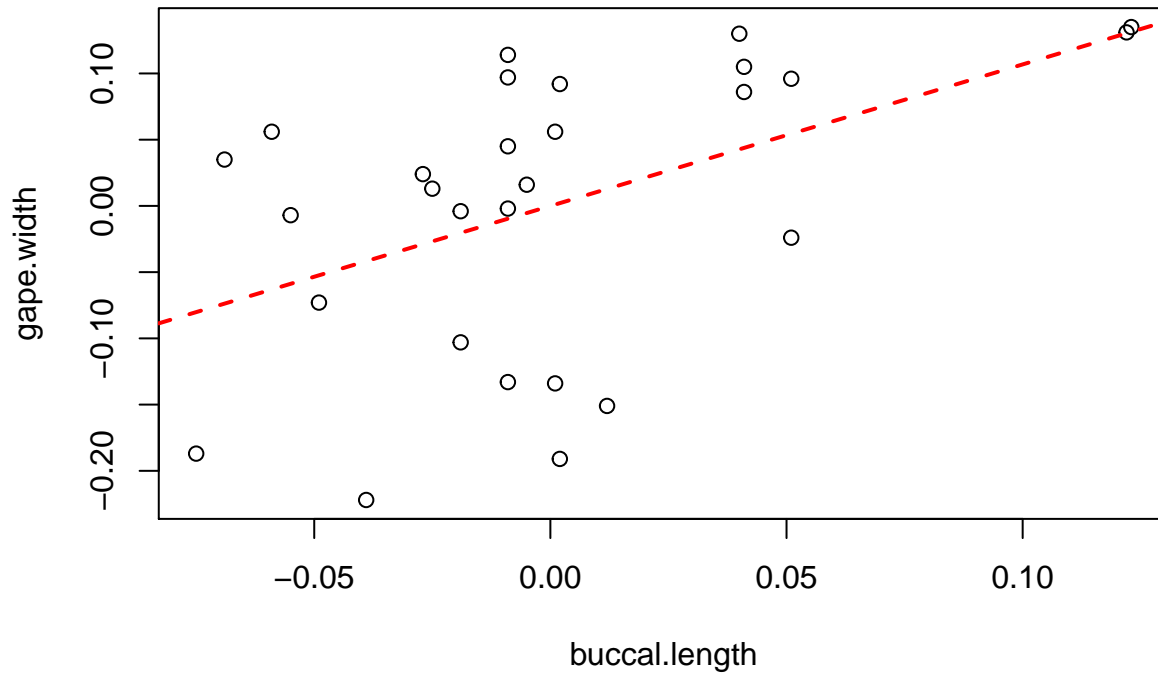
```
##
## Call:
## lm(formula = gape.width ~ buccal.length, data = obj)
##
## Coefficients:
## (Intercept)  buccal.length
##   -3.817e-05    1.069e+00

summary(fit.ols)

##
## Call:
## lm(formula = gape.width ~ buccal.length, data = obj)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.19310 -0.07951  0.03057  0.05653  0.12366
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -3.817e-05  1.832e-02  -0.002  0.99835
## buccal.length  1.069e+00  3.843e-01   2.781  0.00995 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.09695 on 26 degrees of freedom
## Multiple R-squared:  0.2292, Adjusted R-squared:  0.1996
## F-statistic: 7.733 on 1 and 26 DF,  p-value: 0.009951
```

Add OLS regression line to graph:

```
plot(obj[,c("buccal.length", "gape.width")])
abline(fit.ols, lwd=2, lty="dashed", col="red")
```



Fit PIC to take into account phylogeny:

```
cent.tree<-read.tree("Centrarchidae.tre")
buccal.length<-setNames(obj[, "buccal.length"], rownames(obj))
gape.width<-setNames(obj[, "gape.width"], rownames(obj))
pic.bl<-pic(buccal.length, cent.tree) #categorize buccal length by phylogeny
pic.gw<-pic(gape.width, cent.tree) #categorize gape width by phylogeny
fit.pic<-lm(pic.gw~pic.bl+0) #fit linear model with phylogeny considered
fit.pic
```

```
##
## Call:
## lm(formula = pic.gw ~ pic.bl + 0)
##
## Coefficients:
## pic.bl
## 0.5932
```

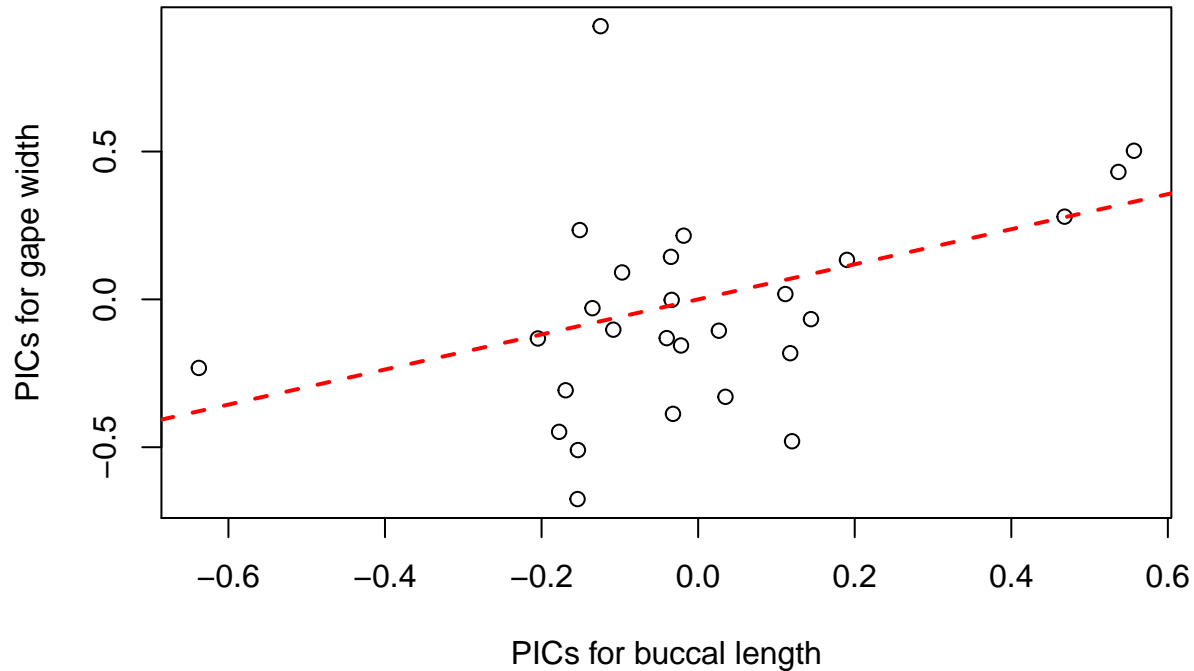
```
summary(fit.pic)
```

```
##
## Call:
## lm(formula = pic.gw ~ pic.bl + 0)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.58401 -0.22945 -0.03814  0.12947  0.99784
##
## Coefficients:
##      Estimate Std. Error t value Pr(>|t|)
## pic.bl    0.5932     0.2556   2.321  0.0284 *
```

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3189 on 26 degrees of freedom
## Multiple R-squared:  0.1716, Adjusted R-squared:  0.1398
## F-statistic: 5.387 on 1 and 26 DF,  p-value: 0.0284
```

Plot linear model over scatterplot with phylogeny considered

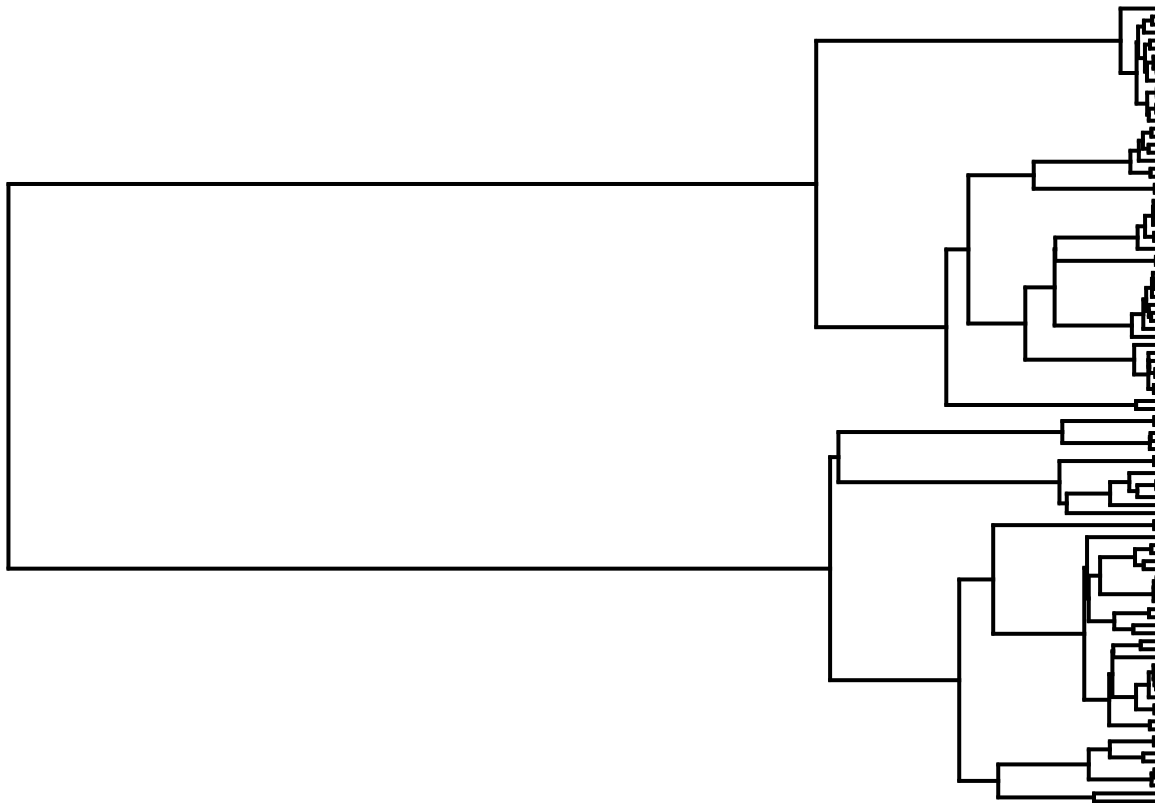
```
plot(pic.bl,pic.gw,xlab="PICs for buccal length",ylab="PICs for gape width")
abline(fit.pic,lwd=2,lty="dashed",col="red")
```



### Simulating a Tree to Test PIC

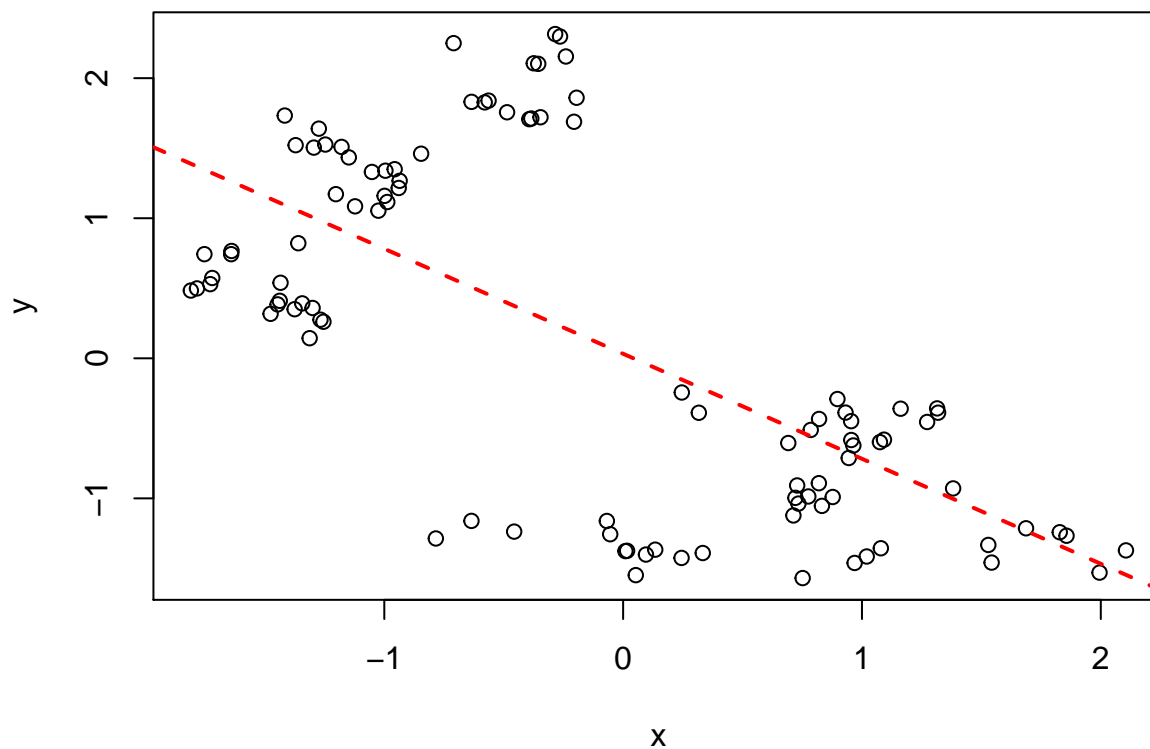
Simulate a coalescent tree:

```
set.seed(21)
## simulate a coalescent shaped tree
tree<-rcoal(n=100)
plotTree(tree,ftype="off")
```



Simulate uncorrelated Brownian evolution:

```
x<-fastBM(tree) #simulate trait 1
y<-fastBM(tree) #simulate trait 2
par(mar=c(5.1,4.1,2.1,2.1)) #set plot dimensions
plot(x,y)
fit<-lm(y~x) #fit linear regression to data
abline(fit,lwd=2,lty="dashed",col="red") #add lm line
```



Return descriptions of fit:

```
fit
```

```
##
## Call:
## lm(formula = y ~ x)
##
## Coefficients:
## (Intercept)          x
##      0.0322      -0.7496
```

```
summary(fit)
```

```
##
## Call:
## lm(formula = y ~ x)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.90636 -0.68689  0.03199  0.53566  2.06929
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.03220    0.09378   0.343   0.732
## x          -0.74963    0.08651  -8.665 9.46e-14 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9294 on 98 degrees of freedom
## Multiple R-squared:  0.4338, Adjusted R-squared:  0.428
## F-statistic: 75.09 on 1 and 98 DF,  p-value: 9.458e-14
```



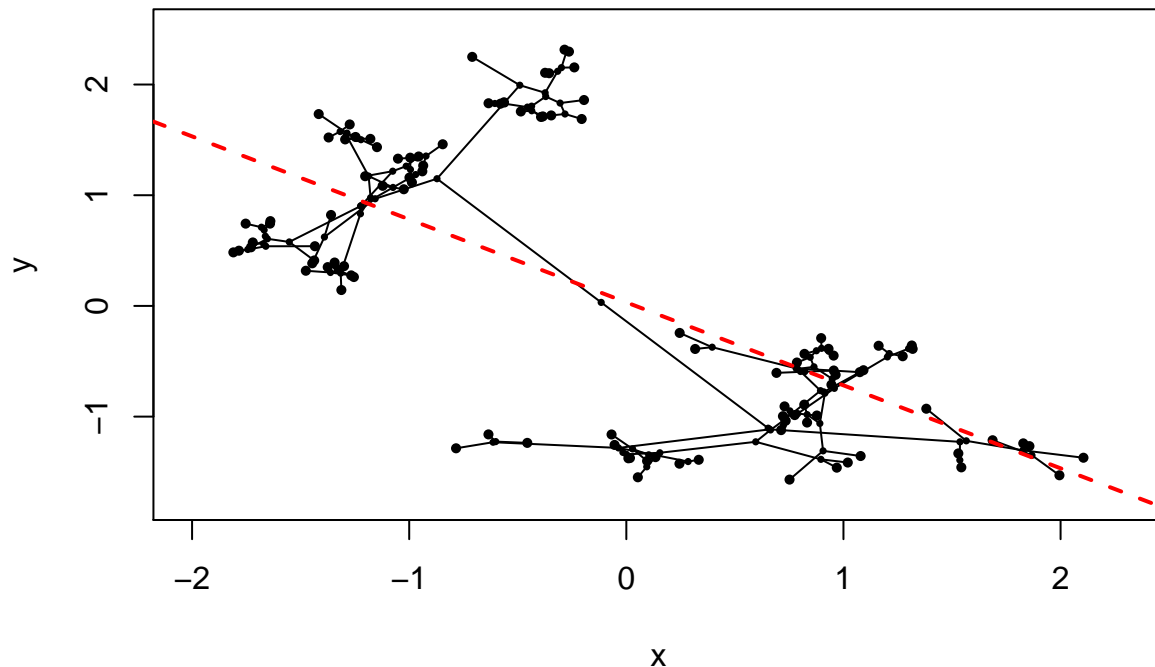
```
anova(fit)
```

```
## Analysis of Variance Table
##
## Response: y
##           Df Sum Sq Mean Sq F value    Pr(>F)
## x           1  64.857   64.857   75.09 9.458e-14 ***
## Residuals  98  84.644    0.864
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Observe that phylogeny can very easily cause a type I error. The close taxonomical relationship of the data points within the two clusters causes the traits to appear correlated even when they are not.

Visualize the relatedness of the data points in the scatterplot:

```
phylomorphospace(tree,cbind(x,y),label="off",node.size=c(0.5,0.7))
abline(fit,lwd=2,lty="dashed",col="red")
```



Use Felsenstein's algorithm to weight the contrasts between species:

```
ix<-pic(x,tree)
iy<-pic(y,tree)
fit<-lm(iy~ix-1) ## we have to fit the model without an intercept term
fit

##
## Call:
## lm(formula = iy ~ ix - 1)
##
## Coefficients:
##          ix
## -0.03771

summary(fit)
```

```
##
## Call:
## lm(formula = iy ~ ix - 1)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.9265 -0.8937 -0.2124  0.4106  2.7002
##
## Coefficients:
##      Estimate Std. Error t value Pr(>|t|)
## ix -0.03771    0.09814  -0.384    0.702
##
## Residual standard error: 1.043 on 98 degrees of freedom
## Multiple R-squared:  0.001504,    Adjusted R-squared:  -0.008685
## F-statistic: 0.1476 on 1 and 98 DF,  p-value: 0.7017
anova(fit)

## Analysis of Variance Table
##
## Response: iy
##           Df Sum Sq Mean Sq F value Pr(>F)
## ix          1   0.16  0.16046   0.1476 0.7017
## Residuals 98 106.53  1.08703
```