

Vertical niches 2.0

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Packages versions:

We used R version 3.3.3 (2017-03-06) and the following packages:

```
##          used (Mb) gc trigger (Mb) max used (Mb)
## Ncells  373459  20.0      592000 31.7    460000 24.6
## Vcells  575161   4.4     1308461 10.0    841783  6.5

##      picante      ape      raster     maptools      rgeos
##      TRUE        TRUE      TRUE        TRUE        TRUE
##      stringr    rgdal    ggplot2      plyr       grid
##      TRUE        TRUE      TRUE        TRUE        TRUE
##      gridExtra multcompView     caper      geiger      phytools
##      TRUE        TRUE      TRUE        TRUE        TRUE
##      knitr       maps     parallel doParallel apTreeshape
##      TRUE        TRUE      TRUE        TRUE        TRUE
##      data.table ppcor    hier.part  relaimpo     MASS
##      TRUE        TRUE      TRUE        TRUE        TRUE
##      bestglm
##      TRUE
```

Load community data

```
# birds
birds <- read.csv("Data/birds_comvars.csv")[,-1]
birds.a <- data.frame(birds[,c(1:2)],
                       log(birds[,c(3:5)]+1),birds[,c(6:11)],
                       log(birds[,c(12:17)]+1),
                       birds[,18])
birds.a <- data.frame(birds[,c(1:2)],scale(birds[,c(3:17)]),birds[,18])
names(birds.a) <- names(birds)

# amphibians
amps <- read.csv("Data/amphibians_comvars.csv")[,-1]
amps <- amps[which(amps$Rich>5),]
amps <- data.frame(amps[c(1:49)])
amps$ses.vert <- amps$ses.vert.realm
amps <- amps[,which(names(amps)%in%names(birds))]
amps <- data.frame(amps[,c(1:2)], amps[,c(3,7:10,12:15)],amps[,c(4:6,16,11)])
amps.a <- data.frame(amps[,c(1:2)], log(amps[,c(3:11)]+1),amps[,c(12:16)])
amps.a <- data.frame(amps.a[,c(1:2)],scale(amps.a[,c(3:15)]),amps.a[,16])
names(amps.a) <- names(amps)

# mammals
mammals <- read.csv("Data/mammals_comvars.csv")[,-1]
mammals.a <- data.frame(mammals[,c(1:2)],
                          log(mammals[,c(3:5)]+1),mammals[,c(6:11)],
                          log(mammals[,c(12:17)]+1),
                          mammals[,18])
mammals.a <- data.frame(mammals[,c(1:2)],scale(mammals[,c(3:17)]),mammals[,18])
names(mammals.a) <- names(mammals)
```

Maps

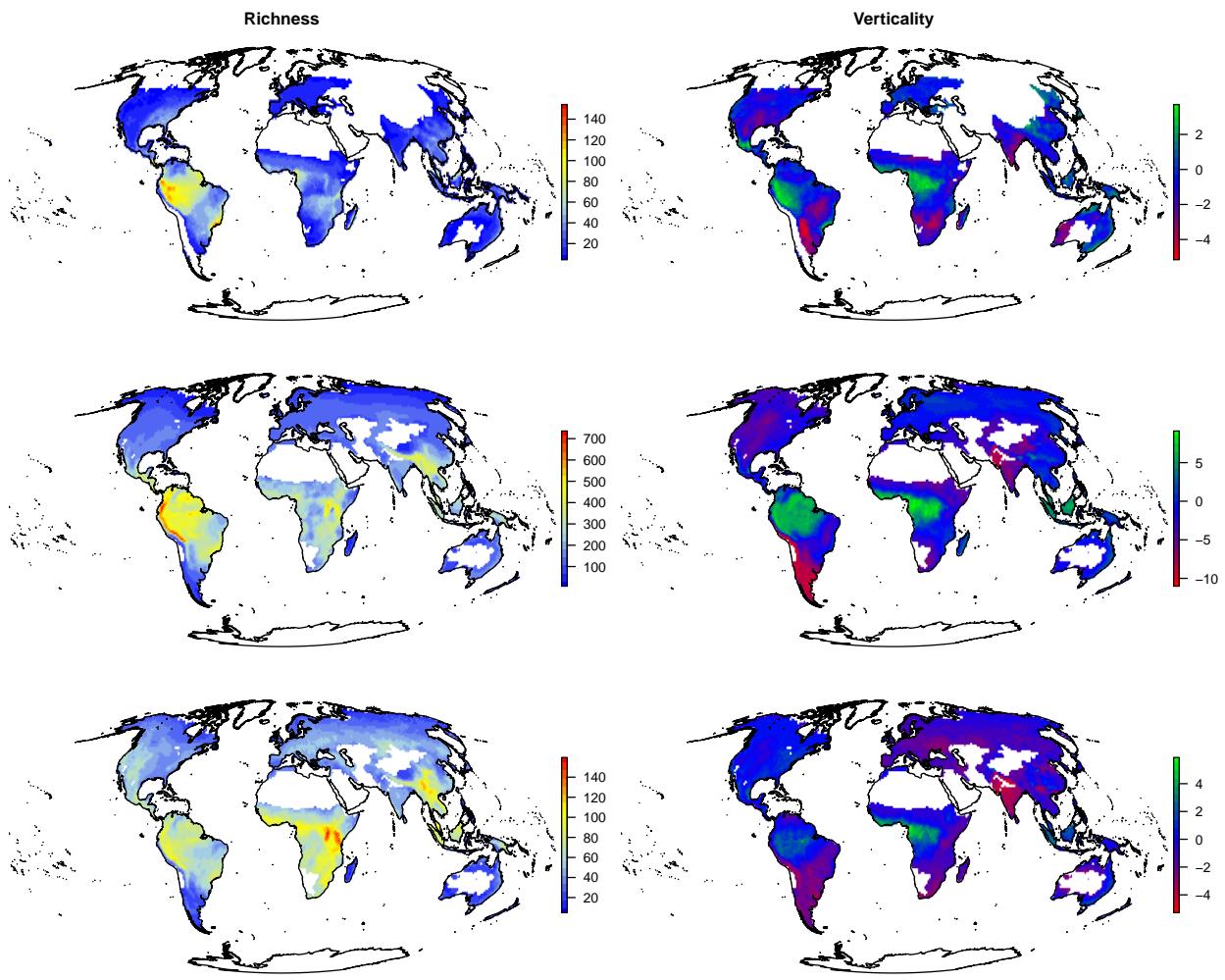
Maps verticality

```
par(mfrow=c(3,2),mar=c(0,0,1,0))

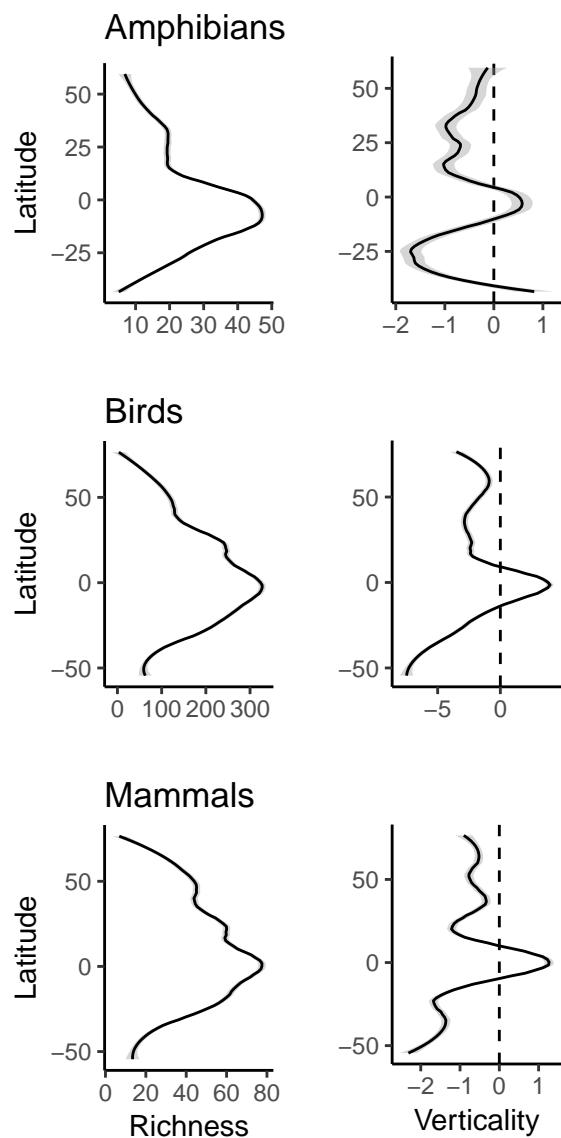
# Amphibians
{plot(s.amps$Rich, main='Richness',axes=F,box=F, col=colorRampPalette(c('blue','lightblue','yellow','red'))(20)
map(mundi,add=T,cex=.5)}
{plot(s.amps$ses.vert, main='Verticality',axes=F,box=F, col=colorRampPalette(c('red','blue','green'))(20)
map(mundi,add=T,cex=.5)}

# Birds
{plot(s.birds$Rich, main=' ',axes=F,box=F, col=colorRampPalette(c('blue','lightblue','yellow','red'))(20)
map(mundi,add=T,cex=.5)}
{plot(s.birds$ses.vert, main=' ',axes=F,box=F, col=colorRampPalette(c('red','blue','green'))(20)
map(mundi,add=T,cex=.5)}

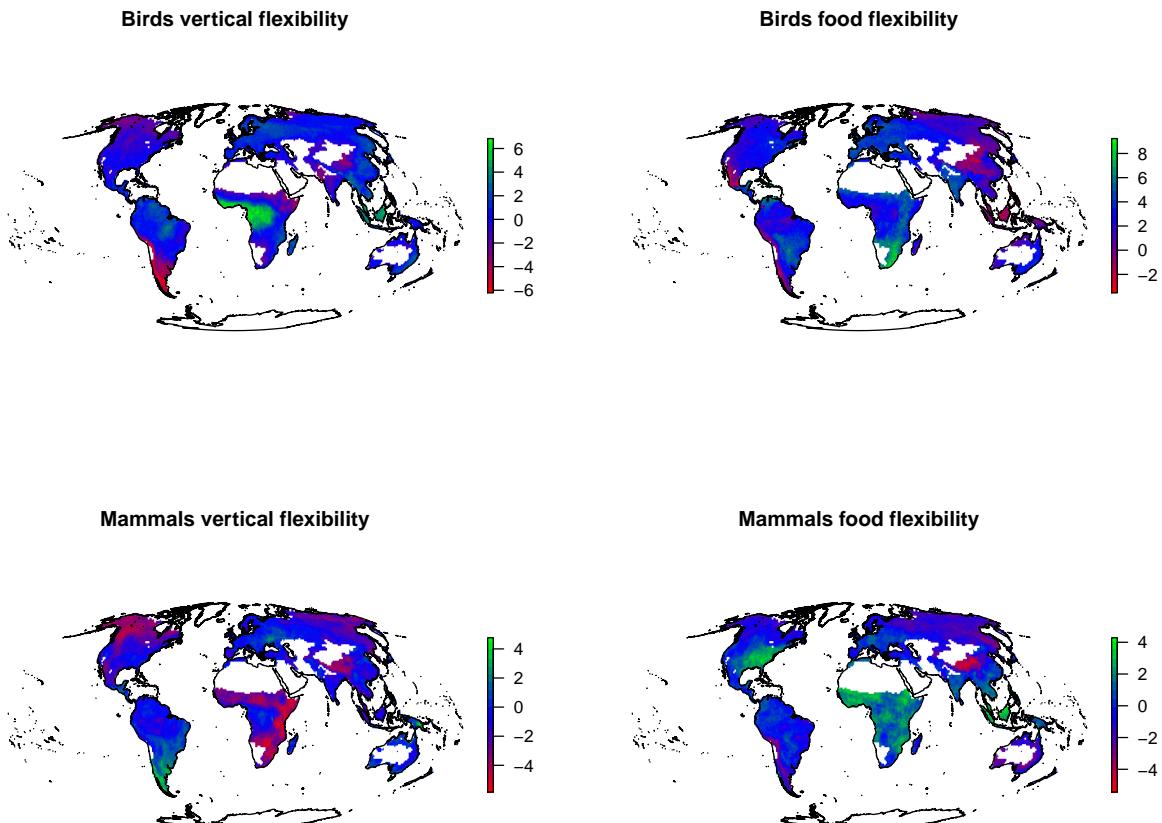
# Mammals
{plot(s.mammals$Rich, main=' ',axes=F,box=F, col=colorRampPalette(c('blue','lightblue','yellow','red'))(20)
map(mundi,add=T,cex=.5)}
{plot(s.mammals$ses.vert, main=' ',axes=F,box=F, col=colorRampPalette(c('red','blue','green'))(20)
map(mundi,add=T,cex=.5)}
```



Latitude



Other metrics



Models

For richness

```
## Amphibians
fit <- glm(Rich ~ ., data = na.omit(amps.a[,-1:-2]))
summary(fit)

##
## Call:
## glm(formula = Rich ~ ., data = na.omit(amps.a[, -1:-2]))
##
## Deviance Residuals:
##      Min        1Q    Median        3Q       Max
## -2.93548  -0.27527   0.02955   0.31120   1.59048
##
## Coefficients:
```

```

##                               Estimate Std. Error t value Pr(>|t|) 
## (Intercept)                 0.081764  0.023007  3.554 0.000382 ***
## velocity                   0.026285  0.010427  2.521 0.011728 *  
## mean.prec                  0.402812  0.013426 30.003 < 2e-16 ***
## sea.prec                   -0.031778  0.009030 -3.519 0.000436 *** 
## veg                        0.163622  0.011860 13.796 < 2e-16 *** 
## elev                        0.270856  0.016203 16.717 < 2e-16 *** 
## WTemp.cell                  -0.174204  0.022688 -7.678 1.84e-14 *** 
## range.size.cell             -0.109391  0.012141 -9.010 < 2e-16 *** 
## body.cell                   0.008846  0.011822  0.748 0.454351 
## mean.temp                   0.390071  0.017315 22.528 < 2e-16 *** 
## sea.temp                     0.116050  0.020904  5.552 2.94e-08 *** 
## diu.temp                     -0.009408  0.011281 -0.834 0.404335 
## ses.vert                     0.027441  0.007942  3.455 0.000553 *** 
## RealmAustralian              -0.416480  0.033906 -12.283 < 2e-16 *** 
## RealmMadagascan              -0.690163  0.080244 -8.601 < 2e-16 *** 
## RealmNearctic                0.071715  0.044757  1.602 0.109127 
## RealmNeotropical              0.372036  0.022219 16.744 < 2e-16 *** 
## RealmOceanina                -1.517958  0.063995 -23.720 < 2e-16 *** 
## RealmOriental                 -0.606926  0.035181 -17.252 < 2e-16 *** 
## RealmPalearctic               -0.187134  0.044127 -4.241 2.26e-05 *** 
## RealmPanamanian                -0.392249  0.052510 -7.470 9.01e-14 *** 
## RealmSaharo-Arabian            -0.421753  0.109015 -3.869 0.000110 *** 
## RealmSino-Japanese              -0.286669  0.051707 -5.544 3.06e-08 *** 
## --- 
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1 
## 
## (Dispersion parameter for gaussian family taken to be 0.247412) 
## 
##      Null deviance: 7028.8 on 6867 degrees of freedom 
## Residual deviance: 1693.5 on 6845 degrees of freedom 
## AIC: 9923 
## 
## Number of Fisher Scoring iterations: 2 
step <- stepAIC(fit, direction="backward") 

## Start:  AIC=9922.96 
## Rich ~ velocity + mean.prec + sea.prec + veg + elev + WTemp.cell + 
##       range.size.cell + body.cell + mean.temp + sea.temp + diu.temp + 
##       ses.vert + Realm 
## 
##                               Df Deviance     AIC 
## - body.cell                 1   1693.7  9921.5 
## - diu.temp                  1   1693.7  9921.7 
## <none>                      1693.5  9923.0 
## - velocity                  1   1695.1  9927.3 
## - ses.vert                   1   1696.5  9932.9 
## - sea.prec                   1   1696.6  9933.4 
## - sea.temp                   1   1701.2  9951.8 
## - WTemp.cell                  1   1708.1  9979.9 
## - range.size.cell             1   1713.6 10001.9 
## - veg                        1   1740.6 10109.3 
## - elev                        1   1762.7 10195.8 
## - mean.temp                   1   1819.1 10412.2

```

```

## - mean.prec      1  1916.3 10769.5
## - Realm         10  2102.1 11387.4
##
## Step: AIC=9921.53
## Rich ~ velocity + mean.prec + sea.prec + veg + elev + WTemp.cell +
##       range.size.cell + mean.temp + sea.temp + diu.temp + ses.vert +
##       Realm
##
##              Df Deviance     AIC
## - diu.temp      1  1693.8 9920.2
## <none>          1693.7 9921.5
## - velocity     1  1695.2 9925.8
## - ses.vert      1  1696.5 9930.9
## - sea.prec      1  1697.0 9933.1
## - sea.temp      1  1701.2 9950.2
## - WTemp.cell    1  1708.2 9978.3
## - range.size.cell 1  1714.2 10002.1
## - veg           1  1741.1 10109.0
## - elev          1  1762.7 10193.8
## - mean.temp     1  1820.2 10414.2
## - mean.prec     1  1916.6 10768.9
## - Realm         10  2136.1 11495.3
##
## Step: AIC=9920.25
## Rich ~ velocity + mean.prec + sea.prec + veg + elev + WTemp.cell +
##       range.size.cell + mean.temp + sea.temp + ses.vert + Realm
##
##              Df Deviance     AIC
## <none>          1693.8 9920.2
## - velocity      1  1695.9 9926.5
## - ses.vert       1  1696.9 9930.7
## - sea.prec       1  1697.7 9933.8
## - sea.temp       1  1701.3 9948.2
## - WTemp.cell     1  1709.3 9980.8
## - range.size.cell 1  1714.2 10000.1
## - veg            1  1741.3 10108.0
## - elev           1  1764.4 10198.4
## - mean.temp      1  1821.1 10415.9
## - mean.prec      1  1935.2 10832.9
## - Realm          10  2178.8 11629.4
step$anova # display results

## Stepwise Model Path
## Analysis of Deviance Table
##
## Initial Model:
## Rich ~ velocity + mean.prec + sea.prec + veg + elev + WTemp.cell +
##       range.size.cell + body.cell + mean.temp + sea.temp + diu.temp +
##       ses.vert + Realm
##
## Final Model:
## Rich ~ velocity + mean.prec + sea.prec + veg + elev + WTemp.cell +
##       range.size.cell + mean.temp + sea.temp + ses.vert + Realm
##

```

```

##          Step Df Deviance Resid. Df Resid. Dev      AIC
## 1                               6845   1693.535 9922.964
## 2 - body.cell  1 0.1385107      6846   1693.674 9921.526
## 3 - diu.temp  1 0.1784845      6847   1693.852 9920.250
## Birds
fit <- glm(Rich ~ ., data = birds.a[,-1:-2])
summary(fit)

## 
## Call:
## glm(formula = Rich ~ ., data = birds.a[, -1:-2])
##
## Deviance Residuals:
##       Min        1Q     Median        3Q       Max
## -1.84167 -0.19731 -0.00927  0.17336  2.88903
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)    
## (Intercept) 0.002996  0.027635  0.108  0.913671    
## WTemp.cell  0.423042  0.030889 13.696 < 2e-16 ***  
## body.cell   -0.030233  0.005588 -5.411 6.41e-08 ***  
## range.size.cell -0.948373  0.019997 -47.426 < 2e-16 ***  
## ses.vert    0.677432  0.010404 65.116 < 2e-16 ***  
## ses.vert.plast -0.240069  0.008685 -27.643 < 2e-16 ***  
## ses.food.plast  0.258363  0.005414 47.725 < 2e-16 ***  
## mean.temp   -0.086812  0.013753 -6.312 2.85e-10 ***  
## sea.temp    -0.097117  0.012534 -7.748 1.01e-14 ***  
## diu.temp    0.064136  0.006284 10.205 < 2e-16 ***  
## mean.prec   0.043123  0.008425  5.119 3.13e-07 ***  
## sea.prec    0.059984  0.005752 10.428 < 2e-16 ***  
## veg         -0.053066  0.006396 -8.297 < 2e-16 ***  
## elev        0.064182  0.004153 15.454 < 2e-16 ***  
## velocity   -0.049966  0.004762 -10.493 < 2e-16 ***  
## RealmAustralian -1.093638  0.025831 -42.338 < 2e-16 ***  
## RealmMadagascan -1.849197  0.057054 -32.411 < 2e-16 ***  
## RealmNearctic -0.190328  0.037498 -5.076 3.92e-07 ***  
## RealmNeotropical  0.486198  0.019348 25.129 < 2e-16 ***  
## RealmOceanina   -0.176942  0.049985 -3.540 0.000402 ***  
## RealmOriental    0.151692  0.029157  5.202 2.00e-07 ***  
## RealmPalaearctic  0.037791  0.044257  0.854 0.393177  
## RealmPanamanian   0.464272  0.040757 11.391 < 2e-16 ***  
## RealmSaharo-Arabian -0.107745  0.048069 -2.241 0.025014 *  
## RealmSino-Japanese -0.160679  0.043766 -3.671 0.000242 ***  
## --- 
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 0.1494052)
##
## Null deviance: 11366.0 on 11366 degrees of freedom
## Residual deviance: 1694.6 on 11342 degrees of freedom
## AIC: 10675
##
## Number of Fisher Scoring iterations: 2

```

```

step <- stepAIC(fit, direction="backward")

## Start: AIC=10675.39
## Rich ~ WTemp.cell + body.cell + range.size.cell + ses.vert +
##       ses.vert.plast + ses.food.plast + mean.temp + sea.temp +
##       diu.temp + mean.prec + sea.prec + veg + elev + velocity +
##       Realm
##
##          Df Deviance    AIC
## <none>           1694.5 10675
## - mean.prec      1   1698.5 10700
## - body.cell       1   1698.9 10703
## - mean.temp       1   1700.5 10713
## - sea.temp        1   1703.5 10733
## - veg             1   1704.8 10742
## - diu.temp        1   1710.1 10777
## - sea.prec        1   1710.8 10782
## - velocity        1   1711.0 10783
## - WTemp.cell       1   1722.6 10860
## - elev            1   1730.2 10910
## - ses.vert.plast  1   1808.7 11414
## - range.size.cell 1   2030.6 12730
## - ses.food.plast  1   2034.8 12754
## - ses.vert         1   2328.0 14284
## - Realm           10  2542.6 15268

```

```
step$anova # display results
```

```

## Stepwise Model Path
## Analysis of Deviance Table
##
## Initial Model:
## Rich ~ WTemp.cell + body.cell + range.size.cell + ses.vert +
##       ses.vert.plast + ses.food.plast + mean.temp + sea.temp +
##       diu.temp + mean.prec + sea.prec + veg + elev + velocity +
##       Realm
##
## Final Model:
## Rich ~ WTemp.cell + body.cell + range.size.cell + ses.vert +
##       ses.vert.plast + ses.food.plast + mean.temp + sea.temp +
##       diu.temp + mean.prec + sea.prec + veg + elev + velocity +
##       Realm
##
##          Step Df Deviance Resid. Df Resid. Dev      AIC
## 1                   11342   1694.553 10675.39
## Mammals
fit <- glm(Rich ~ ., data = mammals.a[,-1:-2])
summary(fit)

##
## Call:
## glm(formula = Rich ~ ., data = mammals.a[, -1:-2])
##
```

```

## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.3133 -0.2628 -0.0068  0.2502  3.7287
##
## Coefficients:
##                               Estimate Std. Error t value Pr(>|t|)
## (Intercept)           0.661774  0.031351 21.108 < 2e-16 ***
## WTemp.cell            0.099613  0.032586  3.057  0.00224 **
## body.cell             -0.191216  0.009763 -19.586 < 2e-16 ***
## range.size.cell       -0.867218  0.020203 -42.925 < 2e-16 ***
## ses.vert              -0.051344  0.007977 -6.436 1.27e-10 ***
## ses.vert.plast        -0.260802  0.009538 -27.345 < 2e-16 ***
## ses.food.plast        -0.026706  0.009065 -2.946  0.00322 **
## mean.temp             -0.078696  0.018738 -4.200 2.69e-05 ***
## sea.temp              -0.105804  0.017118 -6.181 6.60e-10 ***
## diu.temp              -0.062925  0.008119 -7.750 9.94e-15 ***
## mean.prec             0.129443  0.011205 11.552 < 2e-16 ***
## sea.prec              0.013525  0.007797  1.735  0.08284 .
## veg                   0.268975  0.007441 36.146 < 2e-16 ***
## elev                  0.099403  0.005591 17.778 < 2e-16 ***
## velocity              0.011230  0.006357  1.766  0.07734 .
## RealmAustralian      -1.615369  0.040654 -39.735 < 2e-16 ***
## RealmMadagascan       -2.920956  0.081377 -35.894 < 2e-16 ***
## RealmNearctic          -0.935597  0.043568 -21.475 < 2e-16 ***
## RealmNeotropical       -0.994753  0.030746 -32.354 < 2e-16 ***
## RealmOceanina          -2.064709  0.067935 -30.392 < 2e-16 ***
## RealmOriental          -0.644743  0.035571 -18.126 < 2e-16 ***
## RealmPalearctic         -0.462422  0.046586 -9.926 < 2e-16 ***
## RealmPanamanian        -0.761809  0.059507 -12.802 < 2e-16 ***
## RealmSaharo-Arabian    -0.554883  0.055299 -10.034 < 2e-16 ***
## RealmSino-Japanese     -0.845402  0.047366 -17.848 < 2e-16 ***
##
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 0.2695346)
##
## Null deviance: 11293.0 on 11293 degrees of freedom
## Residual deviance: 3037.4 on 11269 degrees of freedom
## AIC: 17271
##
## Number of Fisher Scoring iterations: 2
step <- stepAIC(fit, direction="backward")

## Start:  AIC=17270.86
## Rich ~ WTemp.cell + body.cell + range.size.cell + ses.vert +
##       ses.vert.plast + ses.food.plast + mean.temp + sea.temp +
##       diu.temp + mean.prec + sea.prec + veg + elev + velocity +
##       Realm
##
##                               Df Deviance    AIC
## <none>                      3037.4 17271
## - sea.prec                   1   3038.2 17272
## - velocity                    1   3038.2 17272
## - ses.food.plast               1   3039.7 17278

```

```

## - WTemp.cell      1  3039.9 17278
## - mean.temp      1  3042.1 17287
## - sea.temp       1  3047.7 17307
## - ses.vert        1  3048.6 17310
## - diu.temp        1  3053.6 17329
## - mean.prec       1  3073.4 17402
## - elev            1  3122.6 17581
## - body.cell       1  3140.8 17647
## - ses.vert.plast  1  3238.9 17994
## - veg             1  3389.5 18508
## - range.size.cell 1  3534.0 18979
## - Realm           10 4187.6 20878

step$anova # display results

## Stepwise Model Path
## Analysis of Deviance Table
##
## Initial Model:
## Rich ~ WTemp.cell + body.cell + range.size.cell + ses.vert +
##       ses.vert.plast + ses.food.plast + mean.temp + sea.temp +
##       diu.temp + mean.prec + sea.prec + veg + elev + velocity +
##       Realm
##
## Final Model:
## Rich ~ WTemp.cell + body.cell + range.size.cell + ses.vert +
##       ses.vert.plast + ses.food.plast + mean.temp + sea.temp +
##       diu.temp + mean.prec + sea.prec + veg + elev + velocity +
##       Realm
##
##
##      Step Df Deviance Resid. Df Resid. Dev      AIC
## 1          11269    3037.385 17270.86

```

For verticality

```

## Amphibians
fit <- glm(ses.vert ~ ., data = na.omit(amps.a[,-1:-2]))
summary(fit)

##
## Call:
## glm(formula = ses.vert ~ ., data = na.omit(amps.a[, -1:-2]))
##
## Deviance Residuals:
##      Min      1Q      Median      3Q      Max
## -2.7891 -0.4896 -0.0246  0.4784  3.2770
##
## Coefficients:
##                               Estimate Std. Error t value Pr(>|t|)
## (Intercept)              -0.22590   0.03491 -6.471 1.04e-10 ***
## velocity                  -0.13901   0.01577 -8.813 < 2e-16 ***
## mean.prec                  0.07188   0.02170  3.313 0.000928 ***
## sea.prec                   0.06310   0.01372  4.599 4.33e-06 ***

```

```

## veg          0.37158   0.01772  20.967 < 2e-16 ***
## elev         -0.10044   0.02511  -4.001 6.38e-05 ***
## Rich          0.06345   0.01836   3.455 0.000553 ***
## WTemp.cell    -0.36689   0.03436  -10.677 < 2e-16 ***
## range.size.cell  0.06354   0.01856   3.424 0.000620 ***
## body.cell     -0.27416   0.01767  -15.516 < 2e-16 ***
## mean.temp     -0.13030   0.02724  -4.783 1.76e-06 ***
## sea.temp       0.02176   0.03186   0.683 0.494622
## diu.temp      -0.19730   0.01699  -11.614 < 2e-16 ***
## RealmAustralian  0.63283   0.05156  12.274 < 2e-16 ***
## RealmMadagascan -0.72685   0.12236  -5.940 2.99e-09 ***
## RealmNearctic   0.85451   0.06728  12.701 < 2e-16 ***
## RealmNeotropical -0.45409   0.03403  -13.344 < 2e-16 ***
## RealmOceanina   -0.41317   0.10111  -4.087 4.43e-05 ***
## RealmOriental    -0.12801   0.05462  -2.344 0.019131 *
## RealmPalearctic  0.90018   0.06630  13.577 < 2e-16 ***
## RealmPanamanian  -0.17143   0.08014  -2.139 0.032470 *
## RealmSaharo-Arabian  0.79948   0.16567  4.826 1.42e-06 ***
## RealmSino-Japanese  0.49486   0.07857  6.298 3.20e-10 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 0.5720587)
##
## Null deviance: 6867.0 on 6867 degrees of freedom
## Residual deviance: 3915.7 on 6845 degrees of freedom
## AIC: 15680
##
## Number of Fisher Scoring iterations: 2
step <- stepAIC(fit, direction="backward")

## Start:  AIC=15679.63
## ses.vert ~ velocity + mean.prec + sea.prec + veg + elev + Rich +
##           WTemp.cell + range.size.cell + body.cell + mean.temp + sea.temp +
##           diu.temp + Realm
##
##              Df Deviance   AIC
## - sea.temp      1  3916.0 15678
## <none>            3915.7 15680
## - mean.prec     1  3922.0 15689
## - range.size.cell 1  3922.4 15689
## - Rich          1  3922.6 15690
## - elev          1  3924.9 15694
## - sea.prec       1  3927.8 15699
## - mean.temp      1  3928.8 15700
## - velocity       1  3960.2 15755
## - WTemp.cell     1  3981.0 15791
## - diu.temp       1  3992.9 15812
## - body.cell      1  4053.5 15915
## - veg            1  4167.2 16105
## - Realm          10 4393.2 16450
##
## Step:  AIC=15678.1
## ses.vert ~ velocity + mean.prec + sea.prec + veg + elev + Rich +

```

```

##      WTemp.cell + range.size.cell + body.cell + mean.temp + diu.temp +
##      Realm
##
##              Df Deviance    AIC
## <none>            3916.0 15678
## - mean.prec       1   3922.1 15687
## - range.size.cell 1   3922.8 15688
## - Rich            1   3923.1 15688
## - elev             1   3925.2 15692
## - sea.prec         1   3928.9 15699
## - mean.temp        1   3938.5 15715
## - velocity         1   3969.5 15769
## - WTemp.cell        1   3984.2 15795
## - diu.temp          1   3993.1 15810
## - body.cell         1   4054.3 15914
## - veg              1   4167.6 16104
## - Realm            10  4400.6 16459

step$anova # display results

## Stepwise Model Path
## Analysis of Deviance Table
##
## Initial Model:
## ses.vert ~ velocity + mean.prec + sea.prec + veg + elev + Rich +
##      WTemp.cell + range.size.cell + body.cell + mean.temp + sea.temp +
##      diu.temp + Realm
##
## Final Model:
## ses.vert ~ velocity + mean.prec + sea.prec + veg + elev + Rich +
##      WTemp.cell + range.size.cell + body.cell + mean.temp + diu.temp +
##      Realm
##
##              Step Df  Deviance Resid. Df Resid. Dev      AIC
## 1                      6845   3915.742 15679.63
## 2 - sea.temp  1 0.2668695     6846   3916.009 15678.10

## Birds
fit <- glm(ses.vert ~ ., data = birds.a[,-1:-2])
summary(fit)

##
## Call:
## glm(formula = ses.vert ~ ., data = birds.a[, -1:-2])
##
## Deviance Residuals:
##      Min      1Q      Median      3Q      Max
## -1.80336 -0.17273  0.00219   0.17643   1.66615
##
## Coefficients:
##               Estimate Std. Error t value Pr(>|t|)    
## (Intercept) -0.188808   0.021206 -8.904 < 2e-16 ***
## WTemp.cell   -0.795026   0.022790 -34.886 < 2e-16 ***
## body.cell    -0.020619   0.004304 -4.791 1.68e-06 ***

```

```

## range.size.cell      0.841337  0.014890  56.503 < 2e-16 ***
## ses.vert.plast      0.463147  0.005369  86.268 < 2e-16 ***
## ses.food.plast     -0.213874  0.004103 -52.127 < 2e-16 ***
## mean.temp            0.064276  0.010592  6.068 1.33e-09 ***
## sea.temp              0.075183  0.009652  7.790 7.30e-15 ***
## diu.temp             -0.088919  0.004789 -18.567 < 2e-16 ***
## mean.prec            0.136755  0.006366  21.481 < 2e-16 ***
## sea.prec              0.060585  0.004414  13.726 < 2e-16 ***
## veg                  0.239916  0.004396  54.570 < 2e-16 ***
## elev                 -0.018413  0.003227 -5.706 1.18e-08 ***
## velocity             0.043360  0.003662  11.841 < 2e-16 ***
## Rich                 0.401679  0.006169  65.116 < 2e-16 ***
## RealmAustralian      0.759716  0.020181  37.644 < 2e-16 ***
## RealmMadagascan      0.580167  0.045599  12.723 < 2e-16 ***
## RealmNearctic        0.298747  0.028771  10.384 < 2e-16 ***
## RealmNeotropical     -0.292700  0.015059 -19.437 < 2e-16 ***
## RealmOceanina        -0.720128  0.037913 -18.994 < 2e-16 ***
## RealmOriental         -0.256357  0.022350 -11.470 < 2e-16 ***
## RealmPalearctic       0.388596  0.033884  11.468 < 2e-16 ***
## RealmPanamanian      -0.517188  0.031187 -16.583 < 2e-16 ***
## RealmSaharo-Arabian   0.545435  0.036666  14.876 < 2e-16 ***
## RealmSino-Japanese    0.253740  0.033637   7.543 4.93e-14 ***
##
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 0.08858892)
##
## Null deviance: 11366.0 on 11366 degrees of freedom
## Residual deviance: 1004.8 on 11342 degrees of freedom
## AIC: 4734.4
##
## Number of Fisher Scoring iterations: 2
step <- stepAIC(fit, direction="backward")

## Start:  AIC=4734.37
## ses.vert ~ WTemp.cell + body.cell + range.size.cell + ses.vert.plast +
##          ses.food.plast + mean.temp + sea.temp + diu.temp + mean.prec +
##          sea.prec + veg + elev + velocity + Rich + Realm
##
##          Df Deviance      AIC
## <none>           1004.8  4734.4
## - body.cell       1    1006.8  4755.3
## - elev            1    1007.7  4765.0
## - mean.temp       1    1008.0  4769.2
## - sea.temp         1    1010.1  4793.0
## - velocity        1    1017.2  4872.0
## - sea.prec         1    1021.5  4919.6
## - diu.temp         1    1035.3  5072.7
## - mean.prec        1    1045.7  5185.6
## - WTemp.cell       1    1112.6  5891.0
## - ses.food.plast   1    1245.5  7173.6
## - veg              1    1268.6  7382.4
## - range.size.cell   1    1287.6  7551.6
## - Realm            10   1332.2  7921.0

```

```

## - Rich          1  1380.4  8342.6
## - ses.vert.plast 1  1664.1 10467.0
step$anova # display results

## Stepwise Model Path
## Analysis of Deviance Table
##
## Initial Model:
## ses.vert ~ WTemp.cell + body.cell + range.size.cell + ses.vert.plast +
##           ses.food.plast + mean.temp + sea.temp + diu.temp + mean.prec +
##           sea.prec + veg + elev + velocity + Rich + Realm
##
## Final Model:
## ses.vert ~ WTemp.cell + body.cell + range.size.cell + ses.vert.plast +
##           ses.food.plast + mean.temp + sea.temp + diu.temp + mean.prec +
##           sea.prec + veg + elev + velocity + Rich + Realm
##
##
##   Step Df Deviance Resid. Df Resid. Dev      AIC
## 1           11342    1004.775 4734.371

## Mammals
fit <- glm(ses.vert ~ ., data = mammals.a[,-1:-2])
summary(fit)

##
## Call:
## glm(formula = ses.vert ~ ., data = mammals.a[, -1:-2])
##
## Deviance Residuals:
##    Min      1Q      Median      3Q      Max
## -2.7681  -0.3635  -0.0031   0.3369   3.4985
##
## Coefficients:
##                               Estimate Std. Error t value Pr(>|t|)
## (Intercept)             -0.474464  0.037412 -12.682 < 2e-16 ***
## WTemp.cell              -1.224313  0.036654 -33.402 < 2e-16 ***
## body.cell                -0.002838  0.011702  -0.242   0.808
## range.size.cell         0.178916  0.025632   6.980 3.11e-12 ***
## ses.vert.plast          -0.140135  0.011534 -12.150 < 2e-16 ***
## ses.food.plast          0.054172  0.010677   5.074 3.96e-07 ***
## mean.temp               -0.280354  0.021945 -12.775 < 2e-16 ***
## sea.temp                 0.234794  0.020090  11.687 < 2e-16 ***
## diu.temp                -0.078190  0.009567  -8.173 3.33e-16 ***
## mean.prec                0.341520  0.012891  26.493 < 2e-16 ***
## sea.prec                 0.011640  0.009191   1.266   0.205
## veg                      0.314932  0.008778  35.877 < 2e-16 ***
## elev                     -0.036286  0.006674  -5.437 5.53e-08 ***
## velocity                 -0.002357  0.007494  -0.315   0.753
## Rich                     -0.071337  0.011083  -6.436 1.27e-10 ***
## RealmAustralian          0.417023  0.051016   8.174 3.29e-16 ***
## RealmMadagascan          -1.962856  0.099554 -19.716 < 2e-16 ***
## RealmNearctic            1.605948  0.050163  32.015 < 2e-16 ***
## RealmNeotropical          -0.914925  0.036894 -24.799 < 2e-16 ***

```

```

## RealmOceanina      -1.890655  0.081368 -23.236 < 2e-16 ***
## RealmOriental     -0.659691  0.042079 -15.677 < 2e-16 ***
## RealmPalearctic    0.747974  0.054700 13.674 < 2e-16 ***
## RealmPanamanian   -0.290908  0.070598 -4.121 3.81e-05 ***
## RealmSaharo-Arabian 1.481635  0.063968 23.162 < 2e-16 ***
## RealmSino-Japanese 0.557430  0.056372  9.888 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for gaussian family taken to be 0.3744917)
##
## Null deviance: 11293.0  on 11293  degrees of freedom
## Residual deviance: 4220.1  on 11269  degrees of freedom
## AIC: 20985
##
## Number of Fisher Scoring iterations: 2
step <- stepAIC(fit, direction="backward")

## Start:  AIC=20985.15
## ses.vert ~ WTemp.cell + body.cell + range.size.cell + ses.vert.plast +
##          ses.food.plast + mean.temp + sea.temp + diu.temp + mean.prec +
##          sea.prec + veg + elev + velocity + Rich + Realm
##
##              Df Deviance   AIC
## - body.cell      1  4220.2 20983
## - velocity       1  4220.2 20983
## - sea.prec        1  4220.7 20985
## <none>            4220.1 20985
## - ses.food.plast 1  4229.8 21009
## - elev             1  4231.2 21013
## - Rich             1  4235.7 21025
## - range.size.cell 1  4238.4 21032
## - diu.temp         1  4245.2 21050
## - sea.temp         1  4271.3 21119
## - ses.vert.plast   1  4275.4 21130
## - mean.temp        1  4281.3 21146
## - mean.prec        1  4483.0 21666
## - WTemp.cell       1  4638.0 22049
## - veg              1  4702.2 22205
## - Realm            10 6112.6 25149
##
## Step:  AIC=20983.21
## ses.vert ~ WTemp.cell + range.size.cell + ses.vert.plast + ses.food.plast +
##          mean.temp + sea.temp + diu.temp + mean.prec + sea.prec +
##          veg + elev + velocity + Rich + Realm
##
##              Df Deviance   AIC
## - velocity       1  4220.2 20981
## - sea.prec        1  4220.8 20983
## <none>            4220.2 20983
## - ses.food.plast 1  4229.8 21007
## - elev             1  4231.5 21012
## - Rich             1  4236.0 21024
## - range.size.cell 1  4238.4 21030

```

```

## - diu.temp      1  4245.2 21048
## - sea.temp      1  4272.0 21119
## - mean.temp     1  4281.3 21144
## - ses.vert.plast 1  4303.0 21201
## - mean.prec     1  4484.5 21667
## - WTemp.cell    1  4638.0 22047
## - veg           1  4704.6 22209
## - Realm         10 6202.7 25313
##
## Step: AIC=20981.31
## ses.vert ~ WTemp.cell + range.size.cell + ses.vert.plast + ses.food.plast +
##   mean.temp + sea.temp + diu.temp + mean.prec + sea.prec +
##   veg + elev + Rich + Realm
##
##                               Df Deviance AIC
## - sea.prec      1  4220.8 20981
## <none>          4220.2 20981
## - ses.food.plast 1  4229.8 21005
## - elev          1  4231.7 21010
## - Rich          1  4236.1 21022
## - range.size.cell 1  4238.5 21028
## - diu.temp      1  4245.4 21047
## - sea.temp      1  4273.4 21121
## - mean.temp     1  4286.0 21154
## - ses.vert.plast 1  4305.9 21206
## - mean.prec     1  4484.6 21666
## - WTemp.cell    1  4639.6 22049
## - veg           1  4708.7 22216
## - Realm         10 6323.1 25528
##
## Step: AIC=20980.98
## ses.vert ~ WTemp.cell + range.size.cell + ses.vert.plast + ses.food.plast +
##   mean.temp + sea.temp + diu.temp + mean.prec + veg + elev +
##   Rich + Realm
##
##                               Df Deviance AIC
## <none>          4220.8 20981
## - ses.food.plast 1  4229.9 21003
## - elev          1  4233.0 21012
## - Rich          1  4236.7 21021
## - range.size.cell 1  4238.7 21027
## - diu.temp      1  4246.2 21047
## - sea.temp      1  4277.9 21131
## - mean.temp     1  4287.9 21157
## - ses.vert.plast 1  4321.3 21245
## - mean.prec     1  4484.7 21664
## - WTemp.cell    1  4643.1 22056
## - veg           1  4713.8 22227
## - Realm         10 6739.2 26246
step$anova # display results

```

```

## Stepwise Model Path
## Analysis of Deviance Table
##

```

```

## Initial Model:
## ses.vert ~ WTemp.cell + body.cell + range.size.cell + ses.vert.plast +
##      ses.food.plast + mean.temp + sea.temp + diu.temp + mean.prec +
##      sea.prec + veg + elev + velocity + Rich + Realm
##
## Final Model:
## ses.vert ~ WTemp.cell + range.size.cell + ses.vert.plast + ses.food.plast +
##      mean.temp + sea.temp + diu.temp + mean.prec + veg + elev +
##      Rich + Realm
##
##
##          Step Df    Deviance Resid. Df Resid. Dev      AIC
## 1                      11269   4220.147 20985.15
## 2 - body.cell  1 0.02201893     11270   4220.169 20983.21
## 3 - velocity  1 0.03667512     11271   4220.206 20981.31
## 4 - sea.prec  1 0.62401441     11272   4220.830 20980.98

```

Hierarchical Partitioning

`calc.relimp` calculates several relative importance metrics for the linear model.

For richness

```

## Amphibians
limod <- lm(Rich ~ ., amps.a[,-1:-2])
metrics <- calc.relimp(limod, type = "lmg", rela = T)
as.matrix(100*metrics$lmg[order(100*metrics$lmg, decreasing = T)])

##                               [,1]
## Realm                  21.4773923
## mean.prec              19.8239150
## WTemp.cell              14.3844439
## veg                     11.7318349
## sea.temp                11.3089215
## mean.temp               10.1259373
## elev                    2.3491800
## ses.vert                 1.7677771
## body.cell                1.6853433
## diu.temp                 1.6788406
## range.size.cell          1.4915522
## velocity                 1.4860156
## sea.prec                 0.6888463

## Birds
limod <- lm(Rich ~ ., birds.a[,-1:-2])
metrics <- calc.relimp(limod, type = "lmg", rela = T)
as.matrix(100*metrics$lmg[order(100*metrics$lmg, decreasing = T)])

##                               [,1]
## Realm                  19.9828880
## range.size.cell         12.5464289
## WTemp.cell              10.7163621

```

```

## mean.temp      10.3295253
## sea.temp       9.6315880
## mean.prec      9.2379318
## ses.vert       9.2163741
## veg            5.2015365
## ses.vert.plast 3.6292353
## ses.food.plast 3.1881093
## velocity       1.5048875
## sea.prec        1.4345027
## elev            1.3347295
## diu.temp        1.2702856
## body.cell       0.7756153

## Mammals
limod <- lm(Rich ~ ., mammals.a[,-1:-2])
metrics <- calc.relimp(limod, type = "lmg", rela = T)
as.matrix(100*metrics$lmg[order(100*metrics$lmg, decreasing = T)])

```

```

##                  [,1]
## Realm           24.876115
## range.size.cell 15.442807
## veg             9.990887
## WTemp.cell      8.339483
## mean.temp       7.466124
## sea.temp        6.891017
## ses.food.plast 6.241111
## mean.prec       5.793924
## body.cell       3.357995
## elev            3.335052
## ses.vert.plast 2.933964
## ses.vert         1.380126
## velocity        1.368338
## diu.temp        1.292688
## sea.prec        1.290368

```

For verticality

```

## Amphibians
limod <- lm(ses.vert ~ ., amps.a[,-1:-2])
metrics <- calc.relimp(limod, type = "lmg", rela = T)
as.matrix(100*metrics$lmg[order(100*metrics$lmg, decreasing = T)])

##                  [,1]
## veg              25.160276
## Realm           20.848356
## diu.temp        12.143747
## mean.prec       9.779868
## WTemp.cell      6.220068
## body.cell       5.857209
## Rich            4.202255
## mean.temp       3.798613
## velocity        3.649180
## sea.prec        2.591421
## range.size.cell 2.204848

```

```

## sea.temp      2.103539
## elev         1.440619
## Birds
limod <- lm(ses.vert ~ ., birds.a[,-1:-2])
metrics <- calc.relimp(limod, type = "lmg", rela = T)
as.matrix(100*metrics$lmg[order(100*metrics$lmg, decreasing = T)])
```

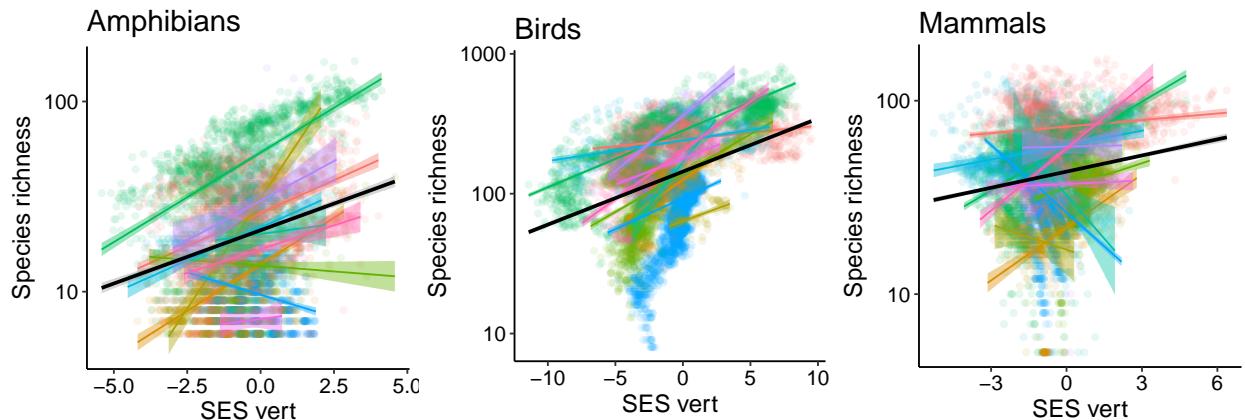
	[,1]
## ses.vert.plast	30.4048072
## veg	17.2247675
## Realm	11.9924505
## Rich	9.4889454
## mean.prec	9.3918452
## diu.temp	4.0337895
## body.cell	4.0136917
## WTemp.cell	3.4924206
## range.size.cell	3.3561556
## ses.food.plast	2.2164929
## mean.temp	1.5132580
## sea.temp	1.2101981
## sea.prec	1.0634770
## velocity	0.3525302
## elev	0.2451706

```

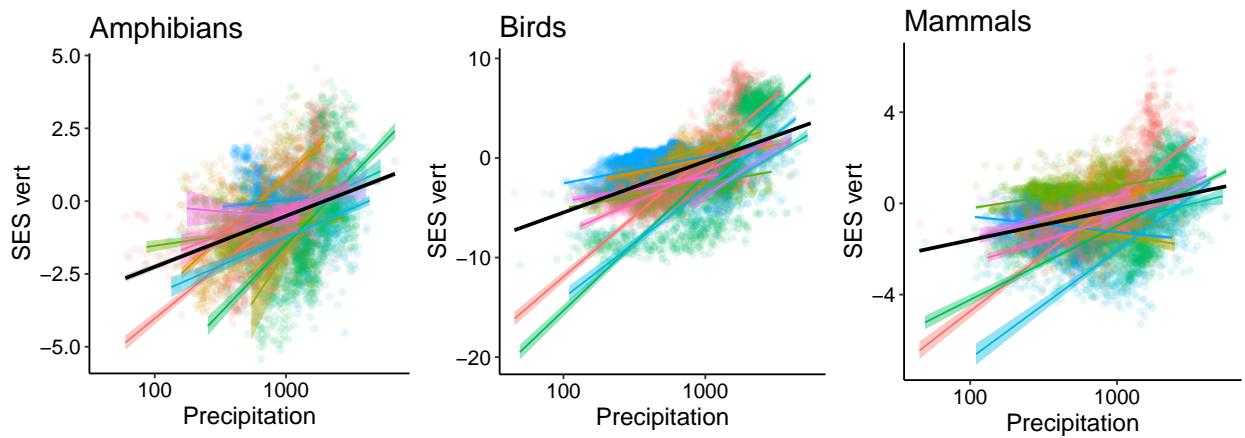
## Mammals
limod <- lm(ses.vert ~ ., mammals.a[,-1:-2])
metrics <- calc.relimp(limod, type = "lmg", rela = T)
as.matrix(100*metrics$lmg[order(100*metrics$lmg, decreasing = T)])
```

	[,1]
## Realm	35.5069868
## veg	17.6865763
## mean.prec	12.2922099
## ses.vert.plast	7.5510296
## WTemp.cell	6.3543712
## sea.prec	4.6607912
## ses.food.plast	4.1086283
## mean.temp	2.0352553
## Rich	1.9975036
## diu.temp	1.5817686
## velocity	1.5195393
## range.size.cell	1.5146348
## sea.temp	1.5017420
## body.cell	1.1404892
## elev	0.5484738

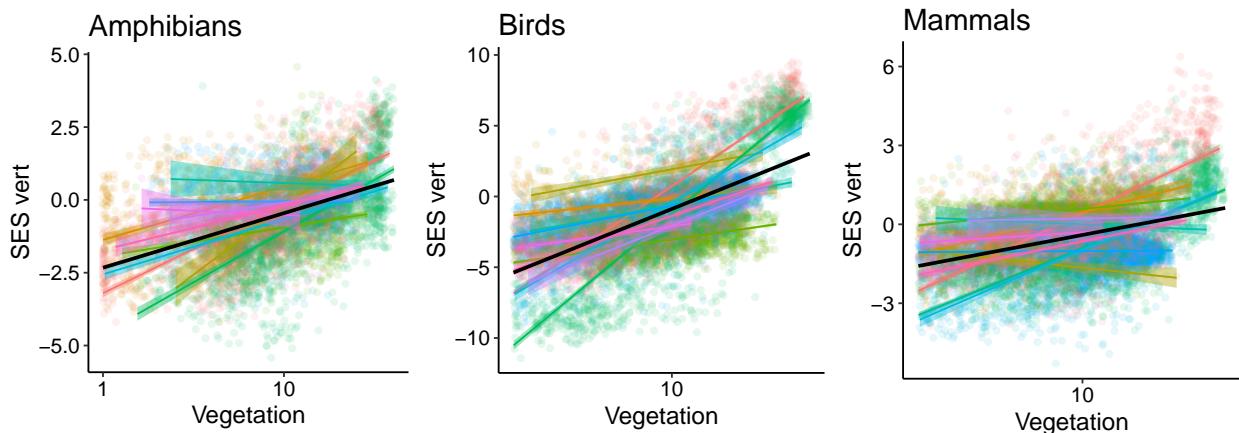
Plot correlations



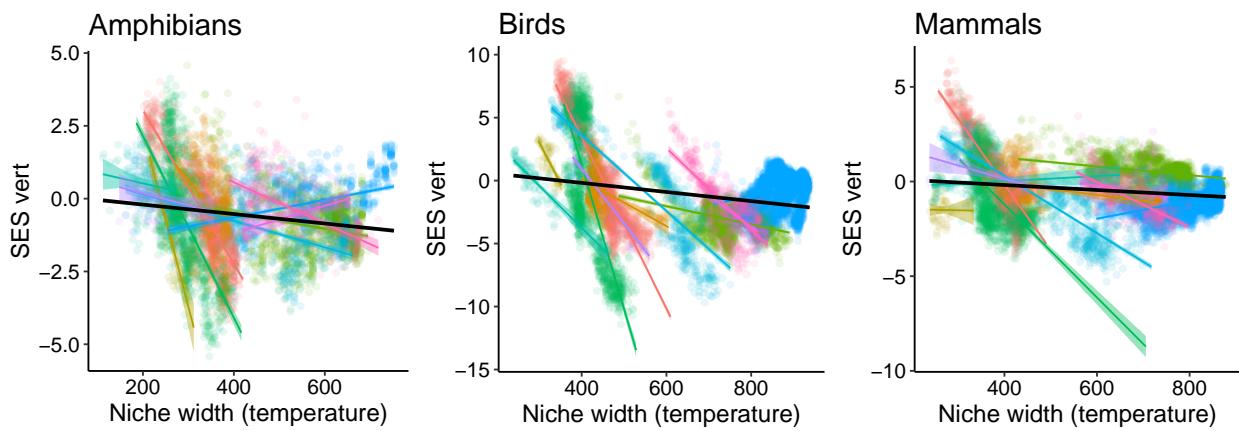
● Afrotropical ● Nearctic ● Oriental ● Saharo-Arabian
 ● Australian ● Neotropical ● Palearctic ● Sino-Japanese
 ● Madagascan ● Oceanina ● Panamanian ● NA



● Afrotropical ● Nearctic ● Oriental ● Saharo-Arabian
 ● Australian ● Neotropical ● Palearctic ● Sino-Japanese
 ● Madagascan ● Oceanina ● Panamanian ● NA



● Afro tropical	● Nearctic	● Oriental	● Saharo-Arabian
● Australian	● Neotropical	● Palearctic	● Sino-Japanese
● Madagascan	● Oceanina	● Panamanian	● NA



● Afro tropical	● Nearctic	● Oriental	● Saharo-Arabian
● Australian	● Neotropical	● Palearctic	● Sino-Japanese
● Madagascan	● Oceanina	● Panamanian	

