Spectra-trait PLSR example using leaf-level spectra and leaf mass per area (LMA) data from 36 species growing in Rosa rugosa invaded coastal grassland communities in Belgium

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Overview

This is an R Markdown Notebook to illustrate how to retrieve a dataset from the EcoSIS spectral database, choose the "optimal" number of plsr components, and fit a plsr model for leaf-mass area (LMA)

Getting Started

Step 1. Load libraries needed to run example script

```
list.of.packages <- c("pls","dplyr","here","plotrix","ggplot2","gridExtra","spectratrait")</pre>
invisible(lapply(list.of.packages, library, character.only = TRUE))
##
## Attaching package: 'pls'
## The following object is masked from 'package:stats':
##
##
       loadings
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
       intersect, setdiff, setequal, union
## here() starts at /Users/sserbin/Data/GitHub/spectratrait
##
## Attaching package: 'gridExtra'
## The following object is masked from 'package:dplyr':
##
##
       combine
```

Step 2. Setup other functions and options

```
### Setup other functions and options
# not in
`%notin%` <- Negate(`%in%`)</pre>
# Script options
pls::pls.options(plsralg = "oscorespls")
pls::pls.options("plsralg")
## $plsralg
## [1] "oscorespls"
# Default par options
opar <- par(no.readonly = T)</pre>
# What is the target variable?
inVar <- "LMA_g_m2"</pre>
# What is the source dataset from EcoSIS?
ecosis_id <- "9db4c5a2-7eac-4e1e-8859-009233648e89"
# Specify output directory, output_dir
# Options:
# tempdir - use a OS-specified temporary directory
# user defined PATH - e.q. "~/scratch/PLSR"
output_dir <- "tempdir"</pre>
```

Step 3. Set working directory (scratch space)

[1] "/private/var/folders/xp/h3k9vf3n2jx181ts786_yjrn9c2gjq/T/RtmpUJeem8"

Step 4. Pull example dataset from EcoSIS (ecosis.org)

```
print(paste0("Output directory: ",getwd())) # check wd
## [1] "Output directory: /Users/sserbin/Data/GitHub/spectratrait/vignettes"
### Get source dataset from EcoSIS
dat_raw <- spectratrait::get_ecosis_data(ecosis_id = ecosis_id)</pre>
## [1] "**** Downloading Ecosis data ****"
## Downloading data...
##
## cols(
##
    .default = col_double(),
    `Latin Species` = col_character(),
##
##
    ids = col_character(),
    `plot code` = col_character(),
    `species code` = col_character()
##
## )
```

```
## i Use `spec()` for the full column specifications.
## Download complete!
head(dat_raw)
## # A tibble: 6 x 2,164
     `Cw/EWT (cm3/cm2)` `Latin Species`
                                               `Leaf area (mm2)` `Leaf calcium cont~
                                                                                <dbl>
##
                  <dbl> <chr>
                                                           <dbl>
## 1
                0.00887 Arrhenatherum elatius
                                                            696.
                                                                              0.0291
## 2
                0.00824 Bromus sterilis
                                                            447.
                                                                              0.0230
## 3
                0.0280 Jacobaea vulgaris
                                                           2418.
                                                                              0.0950
## 4
                0.0106 Rubus caesius
                                                                              0.0700
                                                           5719.
## 5
                0.00851 Arrhenatherum elatius
                                                            671.
                                                                              0.0286
## 6
                0.0153 Crepis capillaris
                                                           1401.
                                                                              0.0470
## # ... with 2,160 more variables:
      Leaf magnesium content per leaf area (mg/mm2) <dbl>,
       Leaf mass per area (g/cm2) <dbl>,
## #
       Leaf nitrogen content per leaf area (mg/mm2) <dbl>,
       Leaf phosphorus content per leaf area (mg/mm2) <dbl>,
## #
       Leaf potassium content per leaf area (mg/mm2) <dbl>,
       Plant height vegetative (cm) <dbl>, ids <chr>, plot code <chr>, ...
names(dat_raw)[1:40]
    [1] "Cw/EWT (cm3/cm2)"
##
   [2] "Latin Species"
##
   [3] "Leaf area (mm2)"
   [4] "Leaf calcium content per leaf area (mg/mm2)"
##
   [5] "Leaf magnesium content per leaf area (mg/mm2)"
   [6] "Leaf mass per area (g/cm2)"
##
  [7] "Leaf nitrogen content per leaf area (mg/mm2)"
   [8] "Leaf phosphorus content per leaf area (mg/mm2)"
   [9] "Leaf potassium content per leaf area (mg/mm2)"
## [10] "Plant height vegetative (cm)"
## [11] "ids"
## [12] "plot code"
## [13] "species code"
## [14] "350"
## [15] "351"
## [16] "352"
## [17] "353"
## [18] "354"
## [19] "355"
## [20] "356"
## [21]
       "357"
## [22] "358"
## [23] "359"
## [24] "360"
## [25]
       "361"
## [26] "362"
## [27] "363"
## [28]
       "364"
## [29] "365"
## [30] "366"
## [31] "367"
```

```
## [32] "368"

## [33] "369"

## [34] "370"

## [35] "371"

## [36] "372"

## [37] "373"

## [38] "374"

## [39] "375"

## [40] "376"
```

Step 5. Create full plsr dataset

```
### Create plsr dataset
Start.wave <- 500
End.wave <- 2400
wv <- seq(Start.wave, End.wave, 1)</pre>
Spectra <- as.matrix(dat_raw[,names(dat_raw) %in% wv])</pre>
colnames(Spectra) <- c(paste0("Wave_",wv))</pre>
sample_info <- dat_raw[,names(dat_raw) %notin% seq(350,2500,1)]</pre>
head(sample_info)
## # A tibble: 6 x 13
##
     `Cw/EWT (cm3/cm2)` `Latin Species`
                                               `Leaf area (mm2)` `Leaf calcium cont~
##
                  <dbl> <chr>
                                                            <dbl>
                                                                                <dbl>
## 1
                0.00887 Arrhenatherum elatius
                                                             696.
                                                                               0.0291
## 2
                0.00824 Bromus sterilis
                                                             447.
                                                                               0.0230
## 3
                0.0280 Jacobaea vulgaris
                                                           2418.
                                                                               0.0950
## 4
                0.0106 Rubus caesius
                                                           5719.
                                                                               0.0700
## 5
                0.00851 Arrhenatherum elatius
                                                            671.
                                                                               0.0286
                0.0153 Crepis capillaris
                                                           1401.
                                                                               0.0470
## # ... with 9 more variables:
      Leaf magnesium content per leaf area (mg/mm2) <dbl>,
## #
       Leaf mass per area (g/cm2) <dbl>,
## #
       Leaf nitrogen content per leaf area (mg/mm2) <dbl>,
## #
       Leaf phosphorus content per leaf area (mg/mm2) <dbl>,
## #
       Leaf potassium content per leaf area (mg/mm2) <dbl>,
       Plant height vegetative (cm) <dbl>, ids <chr>, plot code <chr>, ...
sample_info2 <- sample_info %>%
  select(Plant_Species=`Latin Species`,Species_Code=`species code`,Plot=`plot code`,
         LMA_g_cm2=`Leaf mass per area (g/cm2)`)
sample_info2 <- sample_info2 %>%
  mutate(LMA_g_m2=LMA_g_cm2*10000)
head(sample info2)
## # A tibble: 6 x 5
##
    Plant Species
                           Species_Code Plot LMA_g_cm2 LMA_g_m2
##
     <chr>>
                           <chr>
                                         <chr>
                                                   <dbl>
                                                             <dbl>
## 1 Arrhenatherum elatius Arrela
                                         DC1
                                                 0.00342
                                                              34.2
## 2 Bromus sterilis
                           Broste
                                         DC1
                                                 0.00282
                                                              28.2
## 3 Jacobaea vulgaris
                           Jacvul
                                         DC1
                                                 0.00417
                                                              41.7
                                         DC1
## 4 Rubus caesius
                                                 0.00566
                                                              56.6
                           Rubcae
## 5 Arrhenatherum elatius Arrela
                                         DC2
                                                 0.00361
                                                              36.1
                                         DC2
## 6 Crepis capillaris
                           Creves
                                                 0.00283
                                                              28.3
```

```
plsr_data <- data.frame(sample_info2,Spectra)
rm(sample_info,sample_info2,Spectra)</pre>
```

Step 6. Example data cleaning.

Step 7. Create cal/val datasets

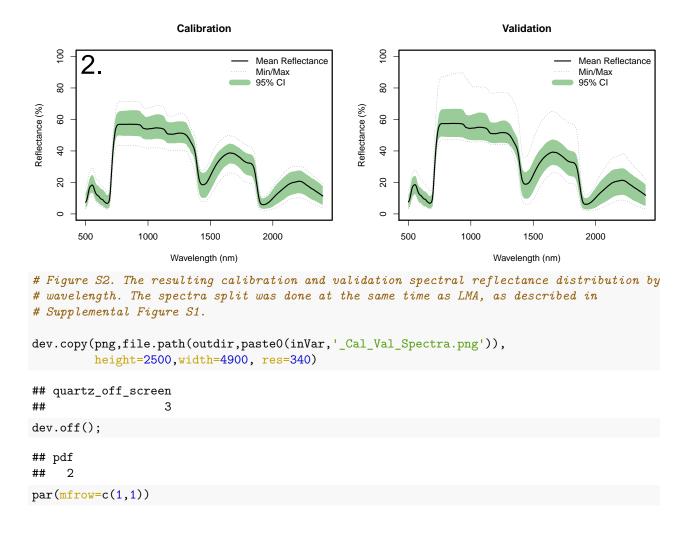
```
method <- "dplyr" #base/dplyr</pre>
\# base R - a bit slow
# dplyr - much faster
split_data <- spectratrait::create_data_split(dataset=plsr_data, approach=method,</pre>
                                              split_seed=7529075, prop=0.8,
                                              group variables="Species Code")
names(split_data)
## [1] "cal data" "val data"
cal.plsr.data <- split_data$cal_data</pre>
head(cal.plsr.data)[1:8]
          Plant_Species Species_Code Plot LMA_g_cm2 LMA_g_m2 Wave_500 Wave_501
##
## 1 Ammophila arenaria
                              Ammare MC2 0.01679492 167.9492 0.135785 0.13685
                              Ammare WC3 0.01844376 184.4376 0.151750 0.15275
## 2 Ammophila arenaria
## 3 Ammophila arenaria
                              Ammare MC4 0.02030190 203.0190 0.156830 0.15790
## 4 Ammophila arenaria
                              Ammare ZC2 0.01591894 159.1894 0.144450 0.14525
                              Ammare ZC1 0.01483469 148.3469 0.147665 0.14910
## 5 Ammophila arenaria
                              Ammare ZC3 0.01802409 180.2409 0.130885 0.13175
## 6 Ammophila arenaria
##
    Wave 502
## 1 0.138150
## 2 0.154150
## 3 0.159065
## 4 0.146220
## 5 0.150330
## 6 0.132750
val.plsr.data <- split_data$val_data</pre>
head(val.plsr.data)[1:8]
##
              Plant_Species Species_Code Plot
                                                LMA_g_cm2 LMA_g_m2 Wave_500
## 1 Arrhenatherum elatius
                                  Arrela DC1 0.003420518 34.20518 0.070667
## 2
            Bromus sterilis
                                  Broste DC1 0.002816940 28.16940 0.105300
## 5 Arrhenatherum elatius
                                  Arrela DC2 0.003611619 36.11619 0.076300
## 6
          Crepis capillaris
                                  Creves DC2 0.002828699 28.28699 0.062717
## 11
                                  Carare DC3 0.010579908 105.79908 0.115885
             Carex arenaria
## 16
           Elytrigia juncea
                                  Elyjun DC4 0.012400353 124.00353 0.116320
##
     Wave_501 Wave_502
```

```
0.07160 0.072533
## 1
## 2
       0.10710 0.109030
       0.07670 0.077300
## 5
       0.06365 0.064850
## 6
## 11 0.11705 0.118450
## 16 0.11745 0.118850
rm(split_data)
# Datasets:
print(paste("Cal observations: ",dim(cal.plsr.data)[1],sep=""))
## [1] "Cal observations: 183"
print(paste("Val observations: ",dim(val.plsr.data)[1],sep=""))
## [1] "Val observations: 73"
text_loc <- c(max(hist(cal.plsr.data[,pasteO(inVar)], plot=FALSE)$counts),</pre>
              max(hist(cal.plsr.data[,paste0(inVar)], plot=FALSE)$mids))
cal_hist_plot <- qplot(cal.plsr.data[,paste0(inVar)],geom="histogram",</pre>
                        main = paste0("Calibration Histogram for ",inVar),
                        xlab = paste0(inVar),ylab = "Count",fill=I("grey50"),col=I("black"),
                        alpha=I(.7)) +
  annotate("text", x=text_loc[2], y=text_loc[1], label= "1.", size=10)
val_hist_plot <- qplot(val.plsr.data[,paste0(inVar)],geom="histogram",</pre>
                        main = paste0("Validation Histogram for ",inVar),
                        xlab = pasteO(inVar),ylab = "Count",fill=I("grey50"),col=I("black"),
                        alpha=I(.7)
histograms <- grid.arrange(cal_hist_plot, val_hist_plot, ncol=2)</pre>
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
   Calibration Histogram for LMA_g_m2
                                                  Validation Histogram for LMA_g_m2
                                           1.
 40 -
Count
                                               Jount 4-
                                                                     80
LMA_g_m2
                     100
LMA_g_m2
# Figure S1. The resulting leaf mass area (LMA, g/m2) distribution (histogram) for the
# calibration (i.e. model training) and validation datasets. The data was split using
# the spectratrait::create_data_split() function using "Species_Code" as the
# group_variable and using a data split proportion per group of 80% to calibration
# and 20% to validation
```

Step 8. Create calibration and validation PLSR datasets

```
### Format PLSR data for model fitting
cal_spec <- as.matrix(cal.plsr.data[, which(names(cal.plsr.data) %in%</pre>
                                              paste0("Wave_",wv))])
cal.plsr.data <- data.frame(cal.plsr.data[, which(names(cal.plsr.data) %notin%</pre>
                                                    paste0("Wave ",wv))],
                            Spectra=I(cal_spec))
head(cal.plsr.data)[1:5]
##
         Plant_Species Species_Code Plot LMA_g_cm2 LMA_g_m2
                              Ammare MC2 0.01679492 167.9492
## 1 Ammophila arenaria
## 2 Ammophila arenaria
                              Ammare WC3 0.01844376 184.4376
## 3 Ammophila arenaria
                              Ammare MC4 0.02030190 203.0190
## 4 Ammophila arenaria
                              Ammare ZC2 0.01591894 159.1894
## 5 Ammophila arenaria
                              Ammare ZC1 0.01483469 148.3469
## 6 Ammophila arenaria
                              Ammare ZC3 0.01802409 180.2409
val_spec <- as.matrix(val.plsr.data[, which(names(val.plsr.data) %in%</pre>
                                              paste0("Wave_",wv))])
val.plsr.data <- data.frame(val.plsr.data[, which(names(val.plsr.data) %notin%</pre>
                                                    paste0("Wave_",wv))],
                            Spectra=I(val_spec))
head(val.plsr.data)[1:5]
##
              Plant_Species_Code Plot LMA_g_cm2 LMA_g_m2
## 1 Arrhenatherum elatius
                                  Arrela DC1 0.003420518 34.20518
           Bromus sterilis
## 2
                                  Broste DC1 0.002816940 28.16940
                                 Arrela DC2 0.003611619 36.11619
## 5 Arrhenatherum elatius
## 6
          Crepis capillaris
                                  Creves DC2 0.002828699 28.28699
## 11
            Carex arenaria
                                  Carare DC3 0.010579908 105.79908
## 16
          Elytrigia juncea
                                  Elyjun DC4 0.012400353 124.00353
```

Step 9. Calibration and Validation spectra plot



Step 10. Use permutation to determine the optimal number of components

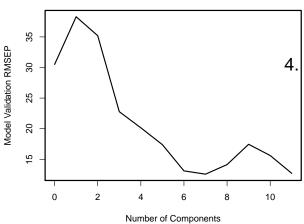
```
### Use permutation to determine the optimal number of components
if(grepl("Windows", sessionInfo()$running)){
 pls.options(parallel = NULL)
} else {
  pls.options(parallel = parallel::detectCores()-1)
method <- "firstMin" #pls, firstPlateau, firstMin</pre>
random_seed <- 7529075
seg <- 80
maxComps <- 16
iterations <- 50
prop <- 0.70
if (method=="pls") {
 nComps <- spectratrait::find_optimal_components(dataset=cal.plsr.data, targetVariable=inVar,</pre>
                                                   method=method,
                                                   maxComps=maxComps, seg=seg,
                                                   random seed=random seed)
  print(paste0("*** Optimal number of components: ", nComps))
```

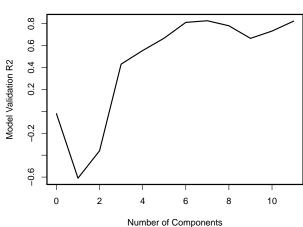
```
} else {
  nComps <- spectratrait::find_optimal_components(dataset=cal.plsr.data, targetVariable=inVar,
                                                   method=method,
                                                   maxComps=maxComps, iterations=iterations,
                                                   seg=seg, prop=prop,
                                                   random_seed=random_seed)
}
## [1] "*** Identifying optimal number of PLSR components ***"
## [1] "*** Running permutation test. Please hang tight, this can take awhile ***"
## [1] "Options:"
## [1] "Max Components: 16 Iterations: 50 Data Proportion (percent): 70"
## [1] "*** Providing PRESS and coefficient array output ***"
## No id variables; using all as measure variables
## [1] "*** Optimal number of components based on t.test: 11"
 60000
 40000
 20000
                                         Number of Components
# Figure S3. Selection of the optimal number of components based on the
# minimization of the PRESS statistic. In this example we show "firstMin"
# option that selects the number of components corresponding to the first
# statistical minimum PRESS value (vertical broken blue line).
dev.copy(png,file.path(outdir,paste0(paste0("Figure_3_",inVar,
                                             "_PLSR_Component_Selection.png"))),
         height=2800, width=3400, res=340)
## quartz_off_screen
##
dev.off();
```

```
## pdf
## 2
```

Step 11. Fit final model

```
### Fit final model - using leave-one-out cross validation
plsr.out <- plsr(as.formula(paste(inVar,"~","Spectra")),scale=FALSE,ncomp=nComps,</pre>
                  validation="L00",trace=FALSE,data=cal.plsr.data)
fit <- plsr.out$fitted.values[,1,nComps]</pre>
pls.options(parallel = NULL)
# External validation fit stats
text_loc <- c(max(RMSEP(plsr.out, newdata = val.plsr.data)$comps),</pre>
              RMSEP(plsr.out, newdata = val.plsr.data)$val[1])
par(mfrow=c(1,2)) \# B, L, T, R
pls::RMSEP(plsr.out, newdata = val.plsr.data)
  (Intercept)
                     1 comps
                                   2 comps
                                                3 comps
                                                              4 comps
                                                                            5 comps
##
         30.50
                       38.30
                                     35.20
                                                  22.78
                                                                              17.39
                                                                20.14
##
       6 comps
                     7 comps
                                   8 comps
                                                9 comps
                                                             10 comps
                                                                           11 comps
                                                  17.45
##
         13.10
                       12.56
                                     14.13
                                                                15.61
                                                                              12.70
plot(pls::RMSEP(plsr.out,estimate=c("test"),newdata = val.plsr.data), main="MODEL RMSEP",
     xlab="Number of Components", ylab="Model Validation RMSEP", lty=1, col="black", cex=1.5, lwd=2)
text(text_loc[1],text_loc[2],labels = "4.", cex=2)
box(1wd=2.2)
pls::R2(plsr.out, newdata = val.plsr.data)
## (Intercept)
                     1 comps
                                   2 comps
                                                3 comps
                                                              4 comps
                                                                            5 comps
##
      -0.02137
                    -0.60981
                                  -0.36001
                                                0.43050
                                                              0.55467
                                                                            0.66818
##
       6 comps
                     7 comps
                                   8 comps
                                                9 comps
                                                             10 comps
                                                                           11 comps
       0.81156
                     0.82673
                                   0.78088
                                                0.66593
                                                              0.73244
                                                                            0.82292
##
plot(pls::R2(plsr.out,estimate=c("test"),newdata = val.plsr.data), main="MODEL R2",
     xlab="Number of Components", ylab="Model Validation R2", lty=1, col="black", cex=1.5, lwd=2)
box(1wd=2.2)
                    MODEL RMSEP
                                                                       MODEL R2
                                                    0.8
                                                    9.0
```





Step 12. PLSR fit observed vs. predicted plot data

```
#calibration
cal.plsr.output <- data.frame(cal.plsr.data[, which(names(cal.plsr.data) %notin%</pre>
                                                       "Spectra")],
                              PLSR Predicted=fit,
                              PLSR_CV_Predicted=as.vector(plsr.out$validation$pred[,,
                                                                                     nComps]))
cal.plsr.output <- cal.plsr.output %>%
  mutate(PLSR_CV_Residuals = PLSR_CV_Predicted-get(inVar))
head(cal.plsr.output)
          Plant_Species Species_Code Plot LMA_g_cm2 LMA_g_m2 PLSR_Predicted
## 1 Ammophila arenaria
                              Ammare MC2 0.01679492 167.9492
                                                                      154.1892
## 2 Ammophila arenaria
                              Ammare WC3 0.01844376 184.4376
                                                                      147.0878
                              Ammare MC4 0.02030190 203.0190
                                                                      153.8674
## 3 Ammophila arenaria
## 4 Ammophila arenaria
                              Ammare ZC2 0.01591894 159.1894
                                                                      161.6047
                              Ammare ZC1 0.01483469 148.3469
## 5 Ammophila arenaria
                                                                      144.9268
                              Ammare ZC3 0.01802409 180.2409
## 6 Ammophila arenaria
                                                                      148,2100
    PLSR CV Predicted PLSR CV Residuals
## 1
                              -16.233027
              151.7161
## 2
              137.3863
                              -47.051273
## 3
              144.2584
                              -58.760574
## 4
              162.6250
                                3.435614
                               -5.436767
## 5
              142.9101
## 6
              142.5160
                              -37.724928
cal.R2 <- round(pls::R2(plsr.out,intercept=F)[[1]][nComps],2)</pre>
cal.RMSEP <- round(sqrt(mean(cal.plsr.output$PLSR_CV_Residuals^2)),2)</pre>
val.plsr.output <- data.frame(val.plsr.data[, which(names(val.plsr.data) %notin%</pre>
                                                        "Spectra")],
                              PLSR_Predicted=as.vector(predict(plsr.out,
                                                                 newdata = val.plsr.data,
                                                                 ncomp=nComps,
                                                                 type="response")[,,1]))
val.plsr.output <- val.plsr.output %>%
 mutate(PLSR Residuals = PLSR Predicted-get(inVar))
```

```
head(val.plsr.output)
                                                LMA_g_cm2 LMA_g_m2 PLSR_Predicted
##
              Plant Species Species Code Plot
## 1
     Arrhenatherum elatius
                                  Arrela DC1 0.003420518 34.20518
                                                                           36.09345
## 2
            Bromus sterilis
                                  Broste DC1 0.002816940 28.16940
                                                                           42.52977
                                  Arrela DC2 0.003611619 36.11619
## 5 Arrhenatherum elatius
                                                                           21.87053
## 6
          Crepis capillaris
                                  Creves DC2 0.002828699 28.28699
                                                                           20.66219
## 11
             Carex arenaria
                                  Carare DC3 0.010579908 105.79908
                                                                           99.79501
## 16
           Elytrigia juncea
                                  Elyjun DC4 0.012400353 124.00353
                                                                          105.16400
##
      PLSR Residuals
## 1
           1.888268
## 2
           14.360370
## 5
         -14.245663
## 6
           -7.624796
## 11
           -6.004066
## 16
          -18.839527
val.R2 <- round(pls::R2(plsr.out, newdata=val.plsr.data, intercept=F)[[1]][nComps],2)
val.RMSEP <- round(sqrt(mean(val.plsr.output$PLSR_Residuals^2)),2)</pre>
rng_quant <- quantile(cal.plsr.output[,inVar], probs = c(0.001, 0.999))</pre>
cal_scatter_plot <- ggplot(cal.plsr.output, aes(x=PLSR_CV_Predicted, y=get(inVar))) +</pre>
  theme_bw() + geom_point() + geom_abline(intercept = 0, slope = 1, color="dark grey",
                                           linetype="dashed", size=1.5) +
  xlim(rng_quant[1], rng_quant[2]) +
  ylim(rng_quant[1], rng_quant[2]) +
  labs(x=paste0("Predicted ", paste(inVar), " (units)"),
       y=paste0("Observed ", paste(inVar), " (units)"),
       title=paste0("Calibration: ", paste0("Rsq = ", cal.R2), "; ",
                    paste0("RMSEP = ", cal.RMSEP))) +
  theme(axis.text=element text(size=18), legend.position="none",
        axis.title=element_text(size=20, face="bold"),
        axis.text.x = element text(angle = 0, vjust = 0.5),
        panel.border = element rect(linetype = "solid", fill = NA, size=1.5)) +
  annotate("text", x=rng quant[1], y=rng quant[2], label= "5.", size=10)
cal_resid_histogram <- ggplot(cal.plsr.output, aes(x=PLSR_CV_Residuals)) +</pre>
  geom_histogram(alpha=.5, position="identity") +
  geom_vline(xintercept = 0, color="black",
             linetype="dashed", size=1) + theme_bw() +
  theme(axis.text=element_text(size=18), legend.position="none",
        axis.title=element_text(size=20, face="bold"),
        axis.text.x = element_text(angle = 0, vjust = 0.5),
        panel.border = element_rect(linetype = "solid", fill = NA, size=1.5))
rng quant <- quantile(val.plsr.output[,inVar], probs = c(0.001, 0.999))
val_scatter_plot <- ggplot(val.plsr.output, aes(x=PLSR_Predicted, y=get(inVar))) +</pre>
  theme_bw() + geom_point() + geom_abline(intercept = 0, slope = 1, color="dark grey",
                                           linetype="dashed", size=1.5) +
  xlim(rng_quant[1], rng_quant[2]) +
  ylim(rng_quant[1], rng_quant[2]) +
  labs(x=paste0("Predicted ", paste(inVar), " (units)"),
```

y=paste0("Observed ", paste(inVar), " (units)"),

title=paste0("Validation: ", paste0("Rsq = ", val.R2), "; ",

```
paste0("RMSEP = ", val.RMSEP))) +
  theme(axis.text=element_text(size=18), legend.position="none",
        axis.title=element_text(size=20, face="bold"),
        axis.text.x = element_text(angle = 0, vjust = 0.5),
        panel.border = element_rect(linetype = "solid", fill = NA, size=1.5))
val_resid_histogram <- ggplot(val.plsr.output, aes(x=PLSR_Residuals)) +</pre>
  geom_histogram(alpha=.5, position="identity") +
  geom_vline(xintercept = 0, color="black",
             linetype="dashed", size=1) + theme_bw() +
  theme(axis.text=element_text(size=18), legend.position="none",
        axis.title=element_text(size=20, face="bold"),
        axis.text.x = element_text(angle = 0, vjust = 0.5),
        panel.border = element_rect(linetype = "solid", fill = NA, size=1.5))
# plot cal/val side-by-side
scatterplots <- grid.arrange(cal_scatter_plot, val_scatter_plot, cal_resid_histogram,</pre>
                             val_resid_histogram, nrow=2, ncol=2)
## Warning: Removed 6 rows containing missing values (geom_point).
## Warning: Removed 3 rows containing missing values (geom_point).
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

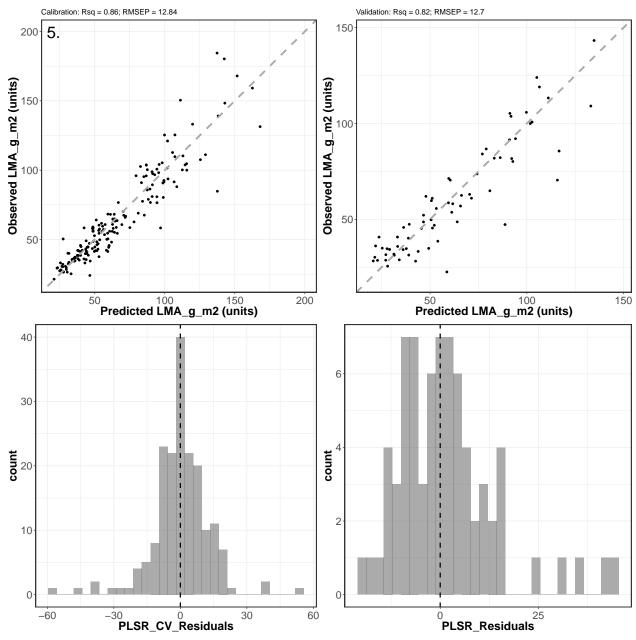
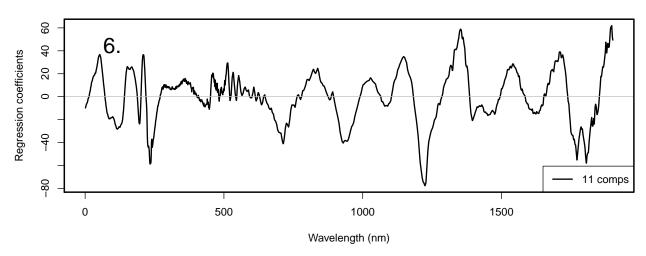


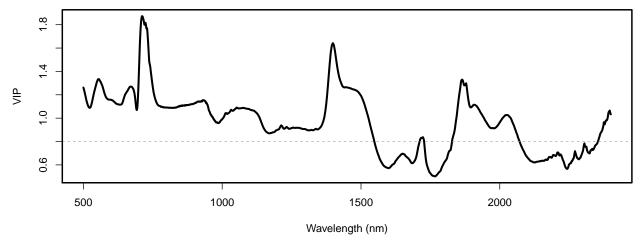
Figure S5. The calibration model and independent validation scatter plot results for # the example LMA PLSR model (top row). Also shown are the calibration model and # validation PLSR residuals, where the calibration results are based on the internal # model cross-validation and the validation residuals are the predicted minus observed # values of LMA.

Step 13. Generate Coefficient and VIP plots

```
ncomp=nComps,lwd=2)
legend("topleft",legend = "6.", cex=2, bty="n")
box(lwd=2.2)
plot(seq(Start.wave,End.wave,1),vips,xlab="Wavelength (nm)",ylab="VIP",cex=0.01)
lines(seq(Start.wave,End.wave,1),vips,lwd=3)
abline(h=0.8,lty=2,col="dark grey")
box(lwd=2.2)
```

LMA_g_m2





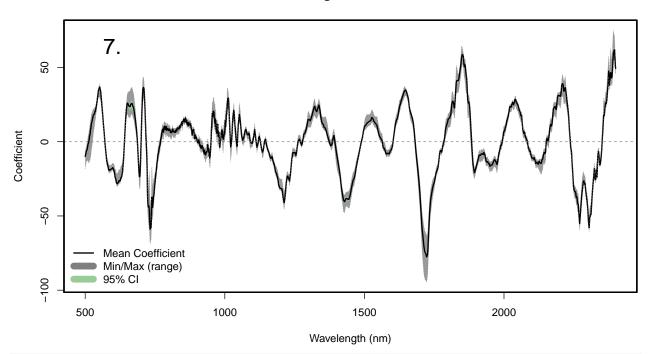
```
## quartz_off_screen
## 3
dev.off();
```

pdf

Step 14. Permutation analysis to derive uncertainty estimates

```
if(grepl("Windows", sessionInfo()$running)){
  pls.options(parallel =NULL)
} else {
  pls.options(parallel = parallel::detectCores()-1)
jk.plsr.out <- pls::plsr(as.formula(paste(inVar,"~","Spectra")), scale=FALSE,
                          center=TRUE, ncomp=nComps, validation="L00", trace=FALSE,
                          jackknife=TRUE,
                          data=cal.plsr.data)
pls.options(parallel = NULL)
Jackknife coef <- spectratrait::f.coef.valid(plsr.out = jk.plsr.out,</pre>
                                              data_plsr = cal.plsr.data,
                                              ncomp = nComps, inVar=inVar)
Jackknife_intercept <- Jackknife_coef[1,,,]</pre>
Jackknife_coef <- Jackknife_coef[2:dim(Jackknife_coef)[1],,,]</pre>
interval \leftarrow c(0.025,0.975)
Jackknife_Pred <- val.plsr.data$Spectra %*% Jackknife_coef +</pre>
  matrix(rep(Jackknife_intercept, length(val.plsr.data[,inVar])), byrow=TRUE,
         ncol=length(Jackknife_intercept))
Interval_Conf <- apply(X = Jackknife_Pred, MARGIN = 1, FUN = quantile,</pre>
                       probs=c(interval[1], interval[2]))
sd_mean <- apply(X = Jackknife_Pred, MARGIN = 1, FUN =sd)</pre>
sd_res <- sd(val.plsr.output$PLSR_Residuals)</pre>
sd_tot <- sqrt(sd_mean^2+sd_res^2)</pre>
val.plsr.output$LCI <- Interval_Conf[1,]</pre>
val.plsr.output$UCI <- Interval_Conf[2,]</pre>
val.plsr.output$LPI <- val.plsr.output$PLSR_Predicted-1.96*sd_tot</pre>
val.plsr.output$UPI <- val.plsr.output$PLSR_Predicted+1.96*sd_tot
head(val.plsr.output)
##
              Plant_Species Species_Code Plot
                                                 LMA_g_cm2 LMA_g_m2 PLSR_Predicted
## 1
     Arrhenatherum elatius
                                   Arrela DC1 0.003420518 34.20518
                                                                             36.09345
## 2
            Bromus sterilis
                                   Broste DC1 0.002816940 28.16940
                                                                             42.52977
## 5
     Arrhenatherum elatius
                                   Arrela DC2 0.003611619 36.11619
                                                                             21.87053
          Crepis capillaris
                                   Creves DC2 0.002828699 28.28699
## 6
                                                                             20.66219
## 11
             Carex arenaria
                                   Carare DC3 0.010579908 105.79908
                                                                             99.79501
## 16
                                   Elyjun DC4 0.012400353 124.00353
                                                                            105.16400
           Elytrigia juncea
      PLSR_Residuals
                                      UCI
                                                LPI
                                                           UPI
##
                           LCI
            1.888268 35.22975 36.83681 11.182998 61.00390
## 1
## 2
           14.360370 41.61622 43.52851 17.617164 67.44238
## 5
          -14.245663 20.07042 23.96996 -3.085793 46.82685
           -7.624796 20.27384 21.15353 -4.234964 45.55935
## 6
## 11
           -6.004066 98.52166 100.58017 74.888636 124.70139
## 16
          -18.839527 104.18470 105.69273 80.260059 130.06795
```

Jackknife regression coefficients



```
# Figure S7. The calibration model jackknife PLSR regression coefficients
dev.copy(png,file.path(outdir,paste0(inVar,'_Jackknife_Regression_Coefficients.png')),
         height=2100, width=3800, res=340)
## quartz_off_screen
dev.off();
## pdf
##
     2
### Permutation validation plot
rmsep_percrmsep <- spectratrait::percent_rmse(plsr_dataset = val.plsr.output,</pre>
                                                 inVar = inVar,
                                                 residuals = val.plsr.output$PLSR_Residuals,
                                                range="full")
RMSEP <- rmsep_percrmsep$rmse
perc_RMSEP <- rmsep_percrmsep$perc_rmse</pre>
r2 <- round(pls::R2(plsr.out, newdata = val.plsr.data,intercept=F)$val[nComps],2)
expr <- vector("expression", 3)</pre>
expr[[1]] \leftarrow bquote(R^2==.(r2))
expr[[2]] <- bquote(RMSEP==.(round(RMSEP,2)))</pre>
expr[[3]] <- bquote("%RMSEP"==.(round(perc_RMSEP,2)))</pre>
```

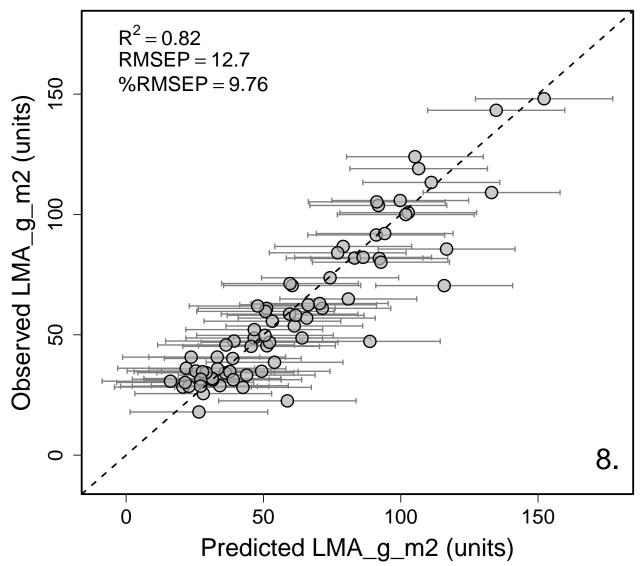


Figure S8. Independent validation results for the LMA PLSR model with associated # jackknife uncertainty estimate 95% prediction intervals for each estimate LMA # value. The %RMSEP is the model prediction performance standardized to the # percentage of the response range, in this case the range of LMA values

Step 15. Output permutation coefficients for later use

```
out.jk.coefs <- data.frame(Iteration=seq(1,length(Jackknife_intercept),1),</pre>
                          Intercept=Jackknife_intercept,t(Jackknife_coef))
head(out.jk.coefs)[1:6]
##
        Iteration Intercept
                              Wave_500 Wave_501
                                                  Wave_502
                                                              Wave_503
## Seg 1
              1 18.33909 -7.580446 -6.724083 -5.886226 -4.984744
                2 21.22164 -8.574931 -7.084795 -6.255716 -5.384000
## Seg 2
## Seg 3
                3 19.63843 -18.104491 -17.260522 -16.154983 -14.960119
## Seg 4
                4 15.90905 -10.715594 -9.874766 -8.926979 -8.007834
## Seg 5
                5 17.51805 -8.952143 -8.305344 -7.136167 -6.221407
## Seg 6
                6 12.18563 -7.702160 -7.128890 -6.532276 -5.840220
write.csv(out.jk.coefs, file=file.path(outdir,
                                     paste0(inVar,
                                            '_Jackkife_PLSR_Coefficients.csv')),
         row.names=FALSE)
```

Step 16. Output remaining core PLSR outputs

```
print(paste("Output directory: ", outdir))
## [1] "Output directory: /var/folders/xp/h3k9vf3n2jx181ts786_yjrn9c2gjq/T//RtmpUJeem8"
# Observed versus predicted
write.csv(cal.plsr.output,file=file.path(outdir,
                                          paste0(inVar,'_Observed_PLSR_CV_Pred_',
                                                 nComps, 'comp.csv')),
          row.names=FALSE)
# Validation data
write.csv(val.plsr.output,file=file.path(outdir,
                                          pasteO(inVar,'_Validation_PLSR_Pred_',
                                                 nComps,'comp.csv')),
          row.names=FALSE)
# Model coefficients
coefs <- coef(plsr.out,ncomp=nComps,intercept=TRUE)</pre>
write.csv(coefs, file=file.path(outdir,
                               pasteO(inVar,'_PLSR_Coefficients_',
                                       nComps,'comp.csv')),
```

Step 17. Confirm files were written to temp space

```
print("**** PLSR output files: ")
## [1] "**** PLSR output files: "
print(list.files(outdir)[grep(pattern = inVar, list.files(outdir))])
    [1] "Figure_3_LMA_g_m2_PLSR_Component_Selection.png"
   [2] "LMA g m2 Cal PLSR Dataset.csv"
##
   [3] "LMA_g_m2_Cal_Val_Histograms.png"
##
   [4] "LMA_g_m2_Cal_Val_Scatterplots.png"
##
##
  [5] "LMA_g_m2_Cal_Val_Spectra.png"
##
  [6] "LMA_g_m2_Coefficient_VIP_plot.png"
  [7] "LMA_g_m2_Jackkife_PLSR_Coefficients.csv"
##
##
   [8] "LMA_g_m2_Jackknife_Regression_Coefficients.png"
  [9] "LMA_g_m2_Observed_PLSR_CV_Pred_11comp.csv"
##
## [10] "LMA_g_m2_PLSR_Coefficients_11comp.csv"
## [11] "LMA_g_m2_PLSR_Validation_Scatterplot.png"
## [12] "LMA_g_m2_PLSR_VIPs_11comp.csv"
## [13] "LMA_g_m2_Val_PLSR_Dataset.csv"
## [14] "LMA_g_m2_Validation_PLSR_Pred_11comp.csv"
## [15] "LMA_g_m2_Validation_RMSEP_R2_by_Component.png"
```