# Spectra-trait PLSR example using leaf-level spectra and leaf mass per area (LMA) data from CONUS NEON sites

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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**

# Load libraries

```
list.of.packages <- c("pls","dplyr","reshape2","here","plotrix","ggplot2","gridExtra",</pre>
                       "spectratrait")
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
```

#### Setup other functions and options

```
### Setup other functions and options
# not in
`%notin%` <- Negate(`%in%`)
# 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_gDW_m2"</pre>
# What is the source dataset from EcoSIS?
ecosis_id <- "5617da17-c925-49fb-b395-45a51291bd2d"
# Specify output directory, output_dir
# Options:
# tempdir - use a OS-specified temporary directory
# user defined PATH - e.q. "~/scratch/PLSR"
output_dir <- "tempdir"</pre>
Set working directory (scratch space)
## [1] "/private/var/folders/xp/h3k9vf3n2jx181ts786_yjrn9c2gjq/T/Rtmpf9WbqL"
Grab data from EcoSIS
print(paste0("Output directory: ",getwd())) # check wd
URL: https://ecosis.org/package/fresh-leaf-spectra-to-estimate-lma-over-neon-domains-in-
eastern-united-states
## [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...
## -- Column specification -----
## cols(
     .default = col_double(),
##
##
    Affiliation = col_character(),
##
   `Common Name` = col_character(),
```

```
##
     Domain = col_character(),
##
     Functional_type = col_character(),
##
     `Latin Genus` = col_character(),
     `Latin Species` = col_character(),
##
##
     PI = col_character(),
     Project = col character(),
##
     Sample ID = col character(),
     `USDA Symbol` = col_character()
##
## )
## i Use `spec()` for the full column specifications.
## Download complete!
head(dat_raw)
## # A tibble: 6 x 2,162
##
     Affiliation
                            `Common Name` Domain Functional type
                                                                    LMA `Latin Genus`
                            <chr>
                                          <chr>
                                                 <chr>>
                                                                  <dbl> <chr>
## 1 University of Wiscon~ black walnut
                                          D02
                                                                   72.9 Juglans
                                                 broadleaf
## 2 University of Wiscon~ black walnut
                                          D02
                                                 broadleaf
                                                                   72.9 Juglans
## 3 University of Wiscon~ black walnut
                                          D02
                                                 broadleaf
                                                                   60.8 Juglans
## 4 University of Wiscon~ black walnut
                                          D02
                                                 broadleaf
                                                                   60.8 Juglans
## 5 University of Wiscon~ black walnut
                                          D02
                                                 broadleaf
                                                                   85.9 Juglans
## 6 University of Wiscon~ black walnut DO2
                                                                   85.9 Juglans
                                                 broadleaf
## # ... with 2,156 more variables: Latin Species <chr>, PI <chr>, Project <chr>,
       Sample_ID <chr>, USDA Symbol <chr>, 350 <dbl>, 351 <dbl>, 352 <dbl>,
## #
       353 <dbl>, 354 <dbl>, 355 <dbl>, 356 <dbl>, 357 <dbl>, 358 <dbl>,
## #
       359 <dbl>, 360 <dbl>, 361 <dbl>, 362 <dbl>, 363 <dbl>, 364 <dbl>,
## #
       365 <dbl>, 366 <dbl>, 367 <dbl>, 368 <dbl>, 369 <dbl>, 370 <dbl>,
## #
       371 <dbl>, 372 <dbl>, 373 <dbl>, 374 <dbl>, 375 <dbl>, 376 <dbl>,
## #
       377 <dbl>, 378 <dbl>, 379 <dbl>, 380 <dbl>, 381 <dbl>, 382 <dbl>,
## #
       383 <dbl>, 384 <dbl>, 385 <dbl>, 386 <dbl>, 387 <dbl>, 388 <dbl>,
       389 <dbl>, 390 <dbl>, 391 <dbl>, 392 <dbl>, 393 <dbl>, 394 <dbl>,
       395 <dbl>, 396 <dbl>, 397 <dbl>, 398 <dbl>, 399 <dbl>, 400 <dbl>,
## #
       401 <dbl>, 402 <dbl>, 403 <dbl>, 404 <dbl>, 405 <dbl>, 406 <dbl>,
## #
## #
       407 <dbl>, 408 <dbl>, 409 <dbl>, 410 <dbl>, 411 <dbl>, 412 <dbl>,
       413 <dbl>, 414 <dbl>, 415 <dbl>, 416 <dbl>, 417 <dbl>, 418 <dbl>,
## #
       419 <dbl>, 420 <dbl>, 421 <dbl>, 422 <dbl>, 423 <dbl>, 424 <dbl>,
## #
       425 <dbl>, 426 <dbl>, 427 <dbl>, 428 <dbl>, 429 <dbl>, 430 <dbl>,
## #
       431 <dbl>, 432 <dbl>, 433 <dbl>, 434 <dbl>, 435 <dbl>, 436 <dbl>,
       437 <dbl>, 438 <dbl>, 439 <dbl>, 440 <dbl>, 441 <dbl>, 442 <dbl>,
## #
       443 <dbl>, 444 <dbl>, ...
names(dat_raw)[1:40]
##
    [1] "Affiliation"
                           "Common Name"
                                             "Domain"
                                                                "Functional_type"
                                                                "PI"
##
    [5] "LMA"
                           "Latin Genus"
                                             "Latin Species"
   [9] "Project"
                           "Sample_ID"
                                             "USDA Symbol"
                                                                "350"
##
                           "352"
                                             "353"
## [13] "351"
                                                                "354"
                                                                "358"
                                             "357"
## [17]
       "355"
                           "356"
## [21] "359"
                           "360"
                                             "361"
                                                                "362"
## [25] "363"
                                             "365"
                                                                "366"
                           "364"
## [29]
        "367"
                           "368"
                                             "369"
                                                                "370"
                           "372"
                                                                "374"
## [33] "371"
                                             "373"
## [37] "375"
                           "376"
                                             "377"
                                                                "378"
```

#### 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 11
   Affiliation
                           `Common Name` Domain Functional_type    LMA `Latin Genus`
                                         <chr> <chr>
                                                                <dbl> <chr>
##
     <chr>>
                           <chr>
## 1 University of Wiscon~ black walnut DO2
                                                broadleaf
                                                                 72.9 Juglans
## 2 University of Wiscon~ black walnut DO2 broadleaf
                                                                72.9 Juglans
## 3 University of Wiscon~ black walnut DO2 broadleaf
                                                                 60.8 Juglans
## 4 University of Wiscon~ black walnut DO2
                                                broadleaf
                                                                 60.8 Juglans
## 5 University of Wiscon~ black walnut DO2
                                                broadleaf
                                                                 85.9 Juglans
## 6 University of Wiscon~ black walnut DO2
                                                broadleaf
                                                                 85.9 Juglans
## # ... with 5 more variables: Latin Species <chr>, PI <chr>, Project <chr>,
      Sample_ID <chr>, USDA Symbol <chr>
sample_info2 <- sample_info %>%
  select(Domain,Functional_type,Sample_ID, USDA_Species_Code=`USDA_Symbol`,LMA_gDW_m2=LMA)
head(sample_info2)
## # A tibble: 6 x 5
    Domain Functional_type Sample_ID USDA_Species_Code LMA_gDW_m2
##
     <chr> <chr>
                           <chr>
                                      <chr>
                                                             <dbl>
                           P0001
                                      JUNI
                                                              72.9
## 1 D02
           broadleaf
## 2 D02
           broadleaf
                           L0001
                                      JUNI
                                                              72.9
## 3 D02 broadleaf
                           P0002
                                      JUNI
                                                              60.8
## 4 DO2 broadleaf
                           L0002
                                      JUNI
                                                              60.8
## 5 D02 broadleaf
                           P0003
                                      JUNI
                                                              85.9
## 6 D02
           broadleaf
                           L0003
                                      JUNI
                                                              85.9
plsr_data <- data.frame(sample_info2,Spectra)</pre>
rm(sample_info, sample_info2, Spectra)
```

# Create cal/val datasets

```
cal.plsr.data <- split_data$cal_data</pre>
head(cal.plsr.data)[1:8]
     Domain Functional_type Sample_ID USDA_Species_Code LMA_gDW_m2
##
                                                                        Wave 500
## 1
        D08
                  broadleaf
                                 L2644
                                                     ACBA
                                                                44.18 0.04170800
## 2
        D08
                  broadleaf
                                 L2646
                                                     ACBA
                                                                41.71 0.05067800
## 3
        D08
                  broadleaf
                                 L2645
                                                     ACBA
                                                                40.66 0.04701700
## 4
        D08
                  broadleaf
                                 P2639
                                                     ACBA
                                                                44.18 0.04125300
## 5
        D03
                  broadleaf
                                 P0614
                                                     ACFL
                                                                52.91 0.03895800
## 6
        D03
                  broadleaf
                                 L0609
                                                     ACFL
                                                                81.67 0.04186169
##
       Wave_501
                  Wave 502
## 1 0.04208700 0.04283700
## 2 0.05087600 0.05153500
## 3 0.04718200 0.04766500
## 4 0.04150300 0.04247100
## 5 0.03915100 0.03956200
## 6 0.04217802 0.04258768
val.plsr.data <- split_data$val_data</pre>
head(val.plsr.data)[1:8]
        Domain Functional_type Sample_ID USDA_Species_Code LMA_gDW_m2 Wave_500
##
## 4923
                      broadleaf
                                    P2462
           D08
                                                        <NA>
                                                                   21.10 0.044964
## 4924
           D08
                      broadleaf
                                    L2462
                                                        SANI
                                                                  100.72 0.068921
           D08
## 4925
                      broadleaf
                                    P2463
                                                        <NA>
                                                                   29.59 0.036254
## 4926
           D08
                      broadleaf
                                    L2463
                                                        SANI
                                                                   96.48 0.051810
## 4927
           אחת
                      broadleaf
                                    P2464
                                                        <NA>
                                                                   31.08 0.056587
## 4928
           D08
                      broadleaf
                                    L2464
                                                        SANI
                                                                   61.40 0.037310
##
        Wave_501 Wave_502
## 4923 0.045854 0.046911
## 4924 0.069633 0.070254
## 4925 0.036999 0.037671
## 4926 0.052113 0.052896
## 4927 0.057006 0.057734
## 4928 0.037223 0.037671
rm(split_data)
# Datasets:
print(paste("Cal observations: ",dim(cal.plsr.data)[1],sep=""))
## [1] "Cal observations: 4922"
print(paste("Val observations: ",dim(val.plsr.data)[1],sep=""))
## [1] "Val observations: 1390"
cal_hist_plot <- qplot(cal.plsr.data[,paste0(inVar)],geom="histogram",</pre>
                        main = paste0("Cal. Histogram for ",inVar),
                        xlab = pasteO(inVar),ylab = "Count",fill=I("grey50"),
                        col=I("black"),alpha=I(.7))
val_hist_plot <- qplot(val.plsr.data[,paste0(inVar)],geom="histogram",</pre>
                        main = pasteO("Val. 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)
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
   Cal. Histogram for LMA_gDW_m2
                                                      Val. Histogram for LMA_gDW_m2
                                                   100 -
 400
Count
                                                  Count
                                                    50 -
 200
                    100 150
LMA_gDW_m2
            50
                                     200
                                                                                       150
ggsave(filename = file.path(outdir,paste0(inVar,"_Cal_Val_Histograms.png")),
       plot = histograms, device="png", width = 30, height = 12, units = "cm",
       dpi = 300)
# output cal/val data
write.csv(cal.plsr.data, file=file.path(outdir,paste0(inVar,'_Cal_PLSR_Dataset.csv')),
           row.names=FALSE)
write.csv(val.plsr.data,file=file.path(outdir,paste0(inVar,'_Val_PLSR_Dataset.csv')),
          row.names=FALSE)
```

#### 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% paste0("Wave_",wv))])</pre>
cal.plsr.data <- data.frame(cal.plsr.data[, which(names(cal.plsr.data) %notin% paste0("Wave_",wv))],</pre>
                              Spectra=I(cal spec))
head(cal.plsr.data)[1:5]
##
     Domain Functional_type Sample_ID USDA_Species_Code LMA_gDW_m2
                   broadleaf
## 1
        D08
                                  L2644
                                                      ACBA
                                                                 44.18
## 2
        D08
                   broadleaf
                                  L2646
                                                      ACBA
                                                                 41.71
                   broadleaf
## 3
        D08
                                  L2645
                                                                 40.66
                                                      ACBA
## 4
        D08
                   broadleaf
                                  P2639
                                                      ACBA
                                                                 44.18
        D03
                   broadleaf
                                                      ACFL
                                                                 52.91
## 5
                                  P0614
                   broadleaf
                                  L0609
                                                                 81.67
## 6
        D03
                                                      ACFL
val_spec <- as.matrix(val.plsr.data[, which(names(val.plsr.data) %in% pasteO("Wave_",wv))])</pre>
val.plsr.data <- data.frame(val.plsr.data[, which(names(val.plsr.data) %notin% paste0("Wave_",wv))],</pre>
                              Spectra=I(val_spec))
head(val.plsr.data)[1:5]
        Domain Functional_type Sample_ID USDA_Species_Code LMA_gDW_m2
##
## 4923
           D08
                      broadleaf
                                     P2462
                                                         <NA>
## 4924
           D08
                      broadleaf
                                     1.2462
                                                                   100.72
                                                         SANT
## 4925
           D08
                      broadleaf
                                     P2463
                                                         <NA>
                                                                    29.59
```

```
## 4926
           D08
                      broadleaf
                                     L2463
                                                         SANI
                                                                    96.48
                      broadleaf
                                     P2464
                                                         <NA>
                                                                    31.08
## 4927
           B00
## 4928
           D08
                      broadleaf
                                                         SANI
                                                                    61.40
                                     L2464
```

### plot cal and val spectra

```
par(mfrow=c(1,2)) # B, L, T, R
spectratrait::f.plot.spec(Z=cal.plsr.data$Spectra,wv=wv,plot_label="Calibration")
spectratrait::f.plot.spec(Z=val.plsr.data$Spectra,wv=wv,plot_label="Validation")
```

Validation

1500

Wavelength (nm)

Mean Reflectance

Min/Max

95% CI

2000

#### Calibration 100 100 Mean Reflectance Min/Max 95% CI 8 80 Reflectance (%) Reflectance (%) 9 9 40 4 20 20 500 1000 1500 2000 500 1000 Wavelength (nm)

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

## pdf
## 2
par(mfrow=c(1,1))
```

# Use Jackknife permutation to determine 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 <- "firstPlateau" #pls, firstPlateau, firstMin
random_seed <- 2356812
seg <- 250
maxComps <- 20
iterations <- 40</pre>
```

```
prop <- 0.70
if (method=="pls") {
 nComps <- spectratrait::find optimal components(dataset=cal.plsr.data, 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, method=method,
                                                 maxComps=maxComps, iterations=iterations,
                                                 seg=seg, prop=prop,
                                                 random_seed=random_seed)
}
## [1] "*** Running permutation test. Please hang tight, this can take awhile ***"
## [1] "Options:"
## [1] "Max Components: 20 Iterations: 40 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: 12"
                     ++++
dev.copy(png,file.path(outdir,paste0(paste0(inVar,"_PLSR_Component_Selection.png"))),
        height=2800, width=3400, res=340)
## quartz off screen
##
dev.off();
## pdf
##
     2
```

#### Fit final model

```
### Fit final model
segs <- 100
plsr.out <- plsr(as.formula(paste(inVar,"~","Spectra")),scale=FALSE,ncomp=nComps,</pre>
```

```
validation="CV",
                  segments=segs, segment.type="interleaved",trace=FALSE,
                  data=cal.plsr.data)
fit <- plsr.out$fitted.values[,1,nComps]</pre>
pls.options(parallel = NULL)
# External validation fit stats
par(mfrow=c(1,2)) # B, L, T, R
pls::RMSEP(plsr.out, newdata = val.plsr.data)
## (Intercept)
                      1 comps
                                    2 comps
                                                  3 comps
                                                                               5 comps
                                                                4 comps
##
        27.155
                      17.610
                                     16.595
                                                   15.483
                                                                 13.235
                                                                                12.374
##
       6 comps
                     7 comps
                                    8 comps
                                                  9 comps
                                                               10 comps
                                                                              11 comps
##
        11.499
                      10.722
                                     10.269
                                                    9.647
                                                                   9.197
                                                                                 9.319
##
      12 comps
         9.515
##
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)
box(1wd=2.2)
pls::R2(plsr.out, newdata = val.plsr.data)
## (Intercept)
                      1 comps
                                    2 comps
                                                                               5 comps
                                                  3 comps
                                                                4 comps
##
     -0.006901
                    0.576543
                                   0.623949
                                                 0.672643
                                                               0.760799
                                                                             0.790906
##
       6 comps
                     7 comps
                                    8 comps
                                                  9 comps
                                                               10 comps
                                                                             11 comps
##
      0.819456
                    0.843031
                                   0.856001
                                                 0.872913
                                                               0.884511
                                                                             0.881406
      12 comps
##
      0.876368
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 R2
                    MODEL RMSEP
                                                      0.8
   25
Model Validation RMSEP
                                                   Model Validation R2
                                                      9.0
   20
                                                      0.4
   15
                                                      0.2
   9
                                                      0.0
```

0

par(opar)

2

6

Number of Components

8

10

12

0

2

4

6

Number of Components

8

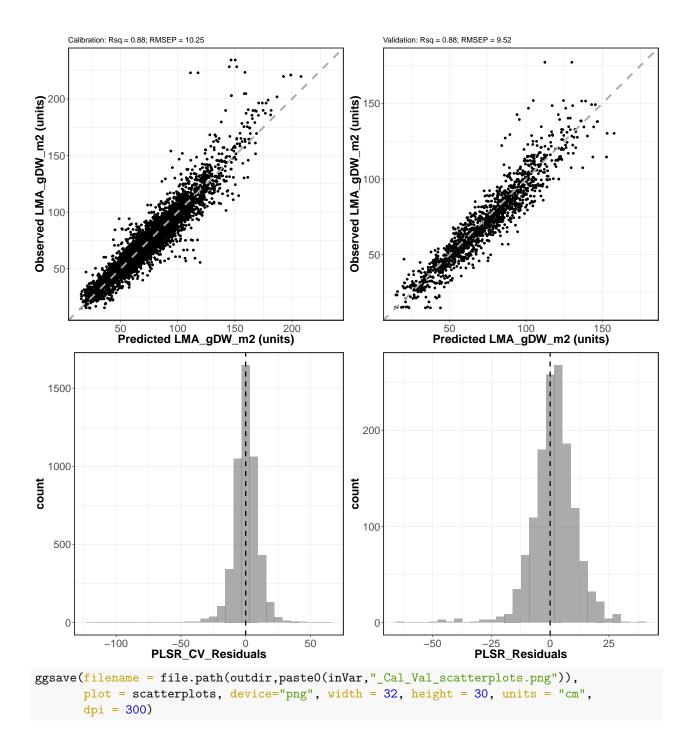
10

12

#### PLSR fit observed vs. predicted plot data

```
#calibration
cal.plsr.output <- data.frame(cal.plsr.data[, which(names(cal.plsr.data) %notin% "Spectra")],</pre>
                               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)
     Domain Functional_type Sample_ID USDA_Species_Code LMA_gDW_m2 PLSR_Predicted
##
## 1
        D08
                  broadleaf
                                 L2644
                                                      ACBA
                                                                44.18
                                                                             53.00773
## 2
        D08
                  broadleaf
                                 L2646
                                                      ACBA
                                                                41.71
                                                                             44.02712
## 3
        D08
                  broadleaf
                                                                             46.76972
                                 L2645
                                                      ACBA
                                                                40.66
## 4
        D08
                  broadleaf
                                 P2639
                                                      ACBA
                                                                44.18
                                                                             49.62804
## 5
        D03
                  broadleaf
                                 P0614
                                                      ACFL
                                                                52.91
                                                                             64.78900
## 6
        D03
                  broadleaf
                                 L0609
                                                      ACFL
                                                                81.67
                                                                             96.85291
     PLSR CV Predicted PLSR CV Residuals
##
## 1
              53.03952
                                 8.859523
## 2
              44.03282
                                 2.322823
              46.90528
## 3
                                 6.245280
## 4
              49.80084
                                 5.620843
## 5
              64.98964
                                12.079641
              96.86039
                                15.190390
cal.R2 <- round(pls::R2(plsr.out,intercept=F)[[1]][nComps],2)
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% "Spectra")],</pre>
                               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)
        Domain Functional_type Sample_ID USDA_Species_Code LMA_gDW_m2
##
## 4923
           D08
                      broadleaf
                                    P2462
                                                         <NA>
                                                                   21.10
## 4924
           D08
                      broadleaf
                                                         SANI
                                                                  100.72
                                    1.2462
## 4925
           D08
                      broadleaf
                                    P2463
                                                         <NA>
                                                                   29.59
## 4926
           D08
                      broadleaf
                                                         SANI
                                                                   96.48
                                    L2463
## 4927
           D08
                      broadleaf
                                    P2464
                                                         <NA>
                                                                   31.08
                                                         SANI
## 4928
           D08
                      broadleaf
                                    L2464
                                                                   61.40
##
        PLSR_Predicted PLSR_Residuals
## 4923
              21.14155
                            0.04155041
              89.65467
## 4924
                          -11.06533484
## 4925
              27.94765
                           -1.64234512
## 4926
              92.46121
                           -4.01879017
## 4927
              40.73367
                            9.65367301
## 4928
              65.94687
                            4.54686556
val.R2 <- round(pls::R2(plsr.out,newdata=val.plsr.data,intercept=F)[[1]][nComps],2)</pre>
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>
```

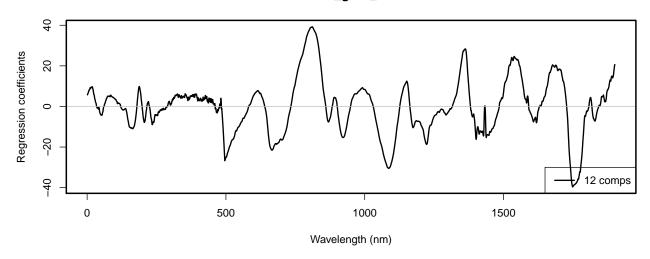
```
cal_scatter_plot <- ggplot(cal.plsr.output, aes(x=PLSR_CV_Predicted,</pre>
                                                 y=get(inVar))) +
  theme_bw() + geom_point() + geom_abline(intercept = 0, slope = 1,
                                           color="dark grey".
                                           linetype="dashed",
                                           size=1.5) +
  xlim(rng_quant[1], rng_quant[2]) +
  vlim(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))
cal_resid_histogram <- ggplot(cal.plsr.output,</pre>
                              aes(x=PLSR_CV_Residuals)) +
  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],</pre>
                      probs = c(0.001, 0.999))
val_scatter_plot <- ggplot(val.plsr.output,</pre>
                           aes(x=PLSR_Predicted, y=get(inVar))) +
  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",
```

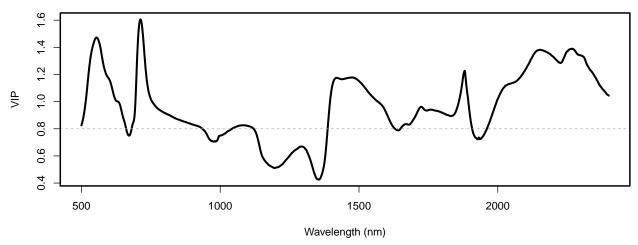


# Generate Coefficient and VIP plots

```
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\_gDW\_m2





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

## pdf ## 2

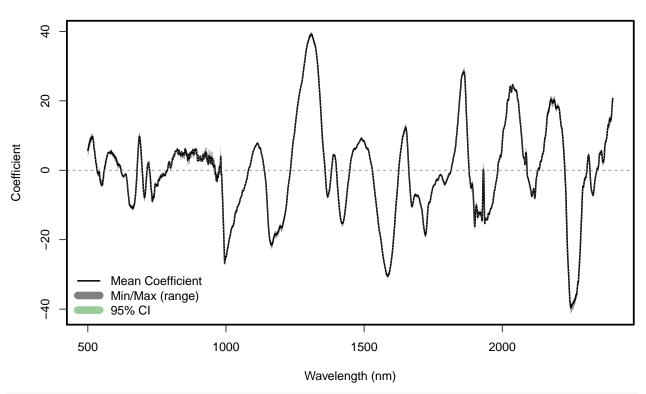
par(opar)

#### Jackknife validation

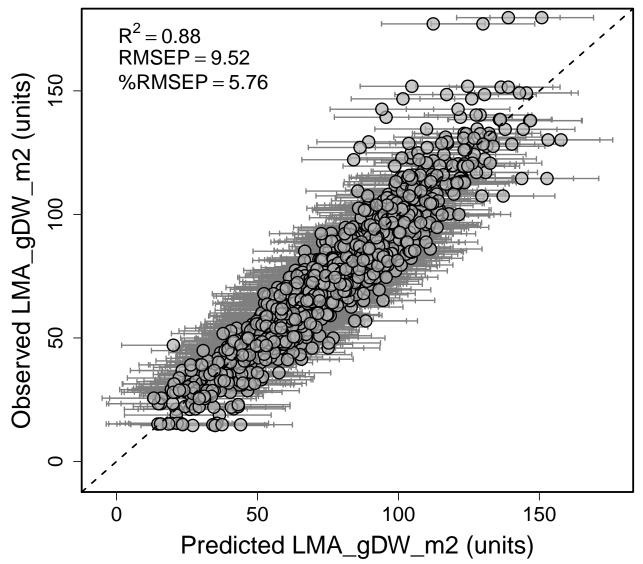
```
if(grepl("Windows", sessionInfo()$running)){
  pls.options(parallel =NULL)
} else {
  pls.options(parallel = parallel::detectCores()-1)
seg <- 100
jk.plsr.out <- pls::plsr(as.formula(paste(inVar,"~","Spectra")), scale=FALSE,
                          center=TRUE, ncomp=nComps,
                          validation="CV", segments = seg,
                          segment.type="interleaved", trace=FALSE,
                          jackknife=TRUE, data=cal.plsr.data)
pls.options(parallel = NULL)
Jackknife_coef <- spectratrait::f.coef.valid(plsr.out = jk.plsr.out, data_plsr = cal.plsr.data,</pre>
                                ncomp = nComps, inVar=inVar)
Jackknife intercept <- Jackknife coef[1,,,]</pre>
Jackknife_coef <- Jackknife_coef[2:dim(Jackknife_coef)[1],,,]</pre>
interval <-c(0.025, 0.975)
Jackknife_Pred <- val.plsr.data$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,</pre>
                        FUN = quantile,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</pre>
head(val.plsr.output)
##
        Domain Functional_type Sample_ID USDA_Species_Code LMA_gDW_m2
## 4923
           D08
                      broadleaf
                                    P2462
                                                         < NA >
                                                                   21.10
## 4924
           D08
                      broadleaf
                                    1.2462
                                                        SANT
                                                                  100.72
## 4925
           D08
                      broadleaf
                                    P2463
                                                         <NA>
                                                                   29.59
## 4926
           D08
                      broadleaf
                                                                   96.48
                                    L2463
                                                         SANI
## 4927
           D08
                      broadleaf
                                    P2464
                                                         <NA>
                                                                   31.08
## 4928
           D08
                      broadleaf
                                    L2464
                                                        SANI
                                                                   61.40
##
        PLSR_Predicted PLSR_Residuals
                                             LCI
                                                      UCI
                                                                 LPI
                                                                            UPI
## 4923
              21.14155
                            0.04155041 20.94847 21.30604 2.789108 39.49399
## 4924
                          -11.06533484 89.45021 89.97246 71.301221 108.00811
              89.65467
## 4925
                          -1.64234512 27.73172 28.16789 9.594597 46.30071
              27.94765
## 4926
              92.46121
                           -4.01879017 92.26489 92.73225 74.107998 110.81442
## 4927
              40.73367
                            9.65367301 40.50065 40.92588 22.380204 59.08714
## 4928
              65.94687
                            4.54686556 65.77618 66.17712 47.594178 84.29955
```

# Jackknife coefficient plot

# Jackknife regression coefficients



# Jackknife validation plot



```
## quartz_off_screen
## 3
dev.off();
## pdf
## 2
```

#### Output jackknife results

```
out.jk.coefs <- data.frame(Iteration=seq(1,seg,1),</pre>
                           Intercept=Jackknife_intercept,
                           t(Jackknife_coef))
head(out.jk.coefs)[1:6]
         Iteration Intercept Wave_500 Wave_501 Wave_502 Wave_503
##
                1 69.44171 5.796399 6.112194 6.482897 6.814518
## Seg 1
                 2 69.36719 6.316698 6.662910 7.031691 7.382632
## Seg 2
## Seg 3
                3 68.73866 6.729135 7.031912 7.372056 7.650510
                4 69.09350 5.899933 6.199538 6.614751 6.975426
## Seg 4
## Seg 5
                5 69.12959 5.355288 5.701339 6.043622 6.383870
                 6 68.67202 5.615024 5.968986 6.358249 6.654510
## Seg 6
write.csv(out.jk.coefs,file=file.path(outdir,
                                      paste0(inVar,
                                             '_Jackkife_PLSR_Coefficients.csv')),
          row.names=FALSE)
```

# Create core PLSR outputs

```
print(paste("Output directory: ", getwd()))
## [1] "Output directory: /Users/sserbin/Data/GitHub/spectratrait/vignettes"
# Observed versus predicted
write.csv(cal.plsr.output,file=file.path(outdir,
                                          pasteO(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')),
          row.names=TRUE)
# PLSR VIP
```

# 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] "LMA_gDW_m2_Cal_PLSR_Dataset.csv"
##
##
   [2] "LMA_gDW_m2_Cal_Val_Histograms.png"
##
   [3] "LMA_gDW_m2_Cal_Val_scatterplots.png"
##
   [4] "LMA_gDW_m2_Cal_Val_Spectra.png"
##
   [5] "LMA gDW m2 Coefficient VIP plot.png"
   [6] "LMA_gDW_m2_Jackkife_PLSR_Coefficients.csv"
##
   [7] "LMA_gDW_m2_Jackknife_Regression_Coefficients.png"
##
## [8] "LMA_gDW_m2_Observed_PLSR_CV_Pred_12comp.csv"
## [9] "LMA_gDW_m2_PLSR_Coefficients_12comp.csv"
## [10] "LMA_gDW_m2_PLSR_Component_Selection.png"
## [11] "LMA gDW m2 PLSR Validation Scatterplot.png"
## [12] "LMA_gDW_m2_PLSR_VIPs_12comp.csv"
## [13] "LMA_gDW_m2_Val_PLSR_Dataset.csv"
## [14] "LMA_gDW_m2_Validation_PLSR_Pred_12comp.csv"
```