An example showing how to apply an existing PLSR model to new data. In this case applying the LMA model from Serbin et al., (2019; DOI - 10.1111/nph.16123) to a dataset collected at CONUS NEON field sites.

Shawn P. Serbin, Julien Lamour, & Jeremiah Anderson

### **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%`)</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_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.g. "~/scratch/PLSR"
output_dir <- "tempdir"</pre>
```

Set working directory (scratch space)

## [1] "/private/var/folders/xp/h3k9vf3n2jx181ts786\_yjrn9c2gjq/T/RtmpOSyBa1"

#### Grab PLSR Coefficients from GitHub

```
git_repo <- "https://raw.githubusercontent.com/serbinsh/SSerbin_etal_2019_NewPhytologist/master/"
print("**** Downloading PLSR coefficients ****")

## [1] "**** Downloading PLSR coefficients ****"
githubURL <- paste0(git_repo, "SSerbin_multibiome_lma_plsr_model/sqrt_LMA_gDW_m2_PLSR_Coefficients_10com_leafLMA.plsr.coeffs <- spectratrait::source_GitHubData(githubURL)
rm(githubURL)
githubURL <- paste0(git_repo, "SSerbin_multibiome_lma_plsr_model/sqrt_LMA_gDW_m2_Jackkife_PLSR_Coefficients_LeafLMA.plsr.jk.coeffs <- spectratrait::source_GitHubData(githubURL)
rm(githubURL)</pre>
```

#### Get source dataset from EcoSIS

```
dat_raw <- spectratrait::get_ecosis_data(ecosis_id = ecosis_id)

## [1] "**** Downloading Ecosis data ****"

## Downloading data...

##

## -- Column specification ------

## cols(

## .default = col_double(),</pre>
```

```
##
     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
                                                 broadleaf
                                                                   72.9 Juglans
## 2 University of Wiscon~ black walnut
                                          D02
                                                 broadleaf
                                                                   72.9 Juglans
## 3 University of Wiscon~ black walnut
                                                                   60.8 Juglans
                                          D02
                                                 broadleaf
                                          D02
## 4 University of Wiscon~ black walnut
                                                                   60.8 Juglans
                                                 broadleaf
## 5 University of Wiscon~ black walnut
                                          D02
                                                 broadleaf
                                                                   85.9 Juglans
## 6 University of Wiscon~ black walnut DO2
                                                 broadleaf
                                                                   85.9 Juglans
     ... 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"
                                                                "350"
##
                                             "USDA Symbol"
## [13] "351"
                           "352"
                                             "353"
                                                                "354"
  [17] "355"
                                             "357"
                                                                "358"
##
                           "356"
                           "360"
                                             "361"
                                                                "362"
##
   [21]
        "359"
## [25]
                           "364"
                                             "365"
                                                                "366"
       "363"
                           "368"
                                             "369"
                                                                "370"
## [29] "367"
## [33] "371"
                           "372"
                                             "373"
                                                                "374"
```

**##** [37] "375" "376" "377" "378"

### Prepare new data for estimation

```
Start.wave <- 500
End.wave <- 2400
wv <- seq(Start.wave, End.wave, 1)</pre>
Spectra <- as.matrix(dat raw[,names(dat raw) %in% wv])
colnames(Spectra) <- c(paste0("Wave_",wv))</pre>
head(Spectra)[1:6,1:10]
        Wave 500 Wave 501 Wave 502 Wave 503 Wave 504 Wave 505 Wave 506 Wave 507
## [1,] 0.044226 0.044605 0.044927 0.045473 0.046241 0.046878 0.047826 0.049090
## [2,] 0.046855 0.047601 0.047944 0.048478 0.049381 0.050235 0.051161 0.052191
## [3,] 0.043758 0.044171 0.044869 0.045465 0.045984 0.046933 0.047993 0.049090
## [4,] 0.041154 0.041603 0.042088 0.042408 0.042639 0.043260 0.044140 0.045058
## [5,] 0.037296 0.037944 0.038209 0.038677 0.039388 0.039948 0.040630 0.041501
## [6,] 0.043878 0.044257 0.044723 0.045295 0.045949 0.046575 0.047378 0.048357
##
        Wave_508 Wave_509
## [1,] 0.050268 0.051525
## [2,] 0.053322 0.054357
## [3,] 0.050168 0.051441
## [4,] 0.045700 0.046476
## [5,] 0.042613 0.043731
## [6,] 0.049392 0.050387
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>>
                                          <chr> <chr>
                                                                  <dbl> <chr>
## 1 University of Wiscon~ black walnut DO2
                                                                  72.9 Juglans
                                                 broadleaf
## 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>
## 1 D02
            broadleaf
                            P0001
                                       JUNI
                                                                72.9
## 2 D02
            broadleaf
                            L0001
                                       JUNI
                                                                72.9
## 3 D02
            broadleaf
                            P0002
                                       JUNI
                                                                60.8
## 4 D02
            broadleaf
                            L0002
                                       JUNI
                                                                60.8
## 5 D02
            broadleaf
                                                               85.9
                            P0003
                                       JUNI
## 6 D02
            broadleaf
                            L0003
                                       JUNI
                                                                85.9
```

```
plsr_data <- data.frame(sample_info2,Spectra)</pre>
rm(sample_info,sample_info2,Spectra)
#### End user needs to do what's appropriate for their data. This may be an iterative process.
# Keep only complete rows of inVar and spec data before fitting
plsr_data <- plsr_data[complete.cases(plsr_data[,names(plsr_data) %in%</pre>
                                                    c(inVar,paste0("Wave_",wv))]),]
Example data cleaning.
print("**** Applying PLSR model to estimate LMA from spectral observations ****")
Prepare PLSR model
## [1] "**** Applying PLSR model to estimate LMA from spectral observations ****"
# setup model
dims <- dim(LeafLMA.plsr.coeffs)</pre>
LeafLMA.plsr.intercept <- LeafLMA.plsr.coeffs[1,]</pre>
LeafLMA.plsr.coeffs <- data.frame(LeafLMA.plsr.coeffs[2:dims[1],])</pre>
names(LeafLMA.plsr.coeffs) <- c("wavelength", "coefs")</pre>
LeafLMA.plsr.coeffs.vec <- as.vector(LeafLMA.plsr.coeffs[,2])</pre>
sub_spec <- droplevels(plsr_data[,which(names(plsr_data) %in%</pre>
                                                     paste0("Wave_",seq(Start.wave,End.wave,1)))])
plsr_pred <- as.matrix(sub_spec) %*% LeafLMA.plsr.coeffs.vec + LeafLMA.plsr.intercept[,2]
leafLMA <- plsr_pred[,1]^2 # convert to standard LMA units from sqrt(LMA)</pre>
names(leafLMA) <- "PLSR_LMA_gDW_m2"</pre>
# organize output
LeafLMA.PLSR.dataset <- data.frame(plsr data[,which(names(plsr data) %notin%
                                                        paste0("Wave_",seq(Start.wave,End.wave,1)))],
                                    PLSR_LMA_gDW_m2=leafLMA, PLSR_Residuals=leafLMA-plsr_data[,inVar])
head(LeafLMA.PLSR.dataset)
Apply PLSR model
     Domain Functional_type Sample_ID USDA_Species_Code LMA_gDW_m2 PLSR_LMA_gDW_m2
## 1
        D02
                  broadleaf
                                 P0001
                                                     JUNI
                                                                72.87
                                                                              96.26243
## 2
        D02
                  broadleaf
                                 L0001
                                                     JUNI
                                                                72.87
                                                                              90.09453
        D02
## 3
                  broadleaf
                                 P0002
                                                     JUNI
                                                                60.77
                                                                             77.16475
## 4
        D02
                  broadleaf
                                 L0002
                                                     JUNI
                                                                60.77
                                                                             60.99039
## 5
        D02
                  broadleaf
                                 P0003
                                                     JUNI
                                                                85.92
                                                                            101.22709
## 6
        D02
                  broadleaf
                                 L0003
                                                     JUNI
                                                                85.92
                                                                             97.13018
    PLSR_Residuals
##
## 1
         23.3924343
## 2
         17.2245326
## 3
         16.3947533
## 4
         0.2203913
## 5
         15.3070857
## 6
         11.2101840
```

```
print("**** Generate PLSR uncertainty estimates ****")
Generate PLSR uncertainty estimates
## [1] "**** Generate PLSR uncertainty estimates ****"
jk coef <- data.frame(LeafLMA.plsr.jk.coeffs[,3:dim(LeafLMA.plsr.jk.coeffs)[2]])</pre>
jk coef <- t(jk coef)</pre>
head(jk_coef)[,1:6]
##
                 [,1]
                            [,2]
                                      [,3]
                                                 [,4]
                                                           [,5]
                                                                      [,6]
## Wave_500 1.0005875 0.9952840 0.5652908 0.9793160 1.1052207 0.9370473
## Wave 501 0.9584235 0.9631434 0.5230544 0.9330803 1.0477469 0.9042780
## Wave_502 0.8960202 0.9065954 0.4597413 0.8710298 0.9658130 0.8628370
## Wave_503 0.8722135 0.8936197 0.4420696 0.8456098 0.9272967 0.8513741
## Wave_504 0.8452831 0.8644923 0.4159567 0.8110004 0.8903192 0.8320347
## Wave_505 0.8240743 0.8378399 0.3902871 0.7829891 0.8570048 0.8150339
jk_int <- t(LeafLMA.plsr.jk.coeffs[,2])</pre>
head(jk_int)[,1:6]
## [1] 7.787098 7.959443 8.015161 8.018586 7.658080 7.998432
jk_pred <- as.matrix(sub_spec) %*% jk_coef + matrix(rep(jk_int, length(plsr_data[,inVar])),</pre>
                                                      byrow=TRUE, ncol=length(jk_int))
jk_pred <- jk_pred^2
head(jk_pred)[,1:6]
         [,1]
                    [,2]
                              [,3]
                                       [,4]
                                                  [,5]
                                                            [,6]
## 1 94.28721 96.77712 96.44452 95.11992 96.72830 95.33877
## 2 90.36051 90.57120 90.77562 89.77821 90.24826 89.61806
## 3 75.71088 77.91861 76.42730 76.11473 77.67179 76.68756
## 4 61.37001 61.30963 60.56606 60.72330 61.63712 60.69649
## 5 99.24456 101.75948 101.22916 99.96305 101.70397 100.16758
## 6 97.40414 97.65463 97.52687 97.00817 97.33677 96.08535
dim(jk_pred)
## [1] 6312 1000
interval <-c(0.025, 0.975)
Interval_Conf <- apply(X = jk_pred, MARGIN = 1, FUN = quantile,</pre>
                       probs=c(interval[1], interval[2]))
sd_mean <- apply(X = jk_pred, MARGIN = 1, FUN =sd)</pre>
sd res <- sd(LeafLMA.PLSR.dataset$PLSR Residuals)</pre>
sd_tot <- sqrt(sd_mean^2+sd_res^2)</pre>
LeafLMA.PLSR.dataset$LCI <- Interval_Conf[1,]</pre>
LeafLMA.PLSR.dataset$UCI <- Interval_Conf[2,]</pre>
LeafLMA.PLSR.dataset$LPI <- LeafLMA.PLSR.dataset$PLSR_LMA_gDW_m2-1.96*sd_tot
LeafLMA.PLSR.dataset$UPI <- LeafLMA.PLSR.dataset$PLSR_LMA_gDW_m2+1.96*sd_tot
head(LeafLMA.PLSR.dataset)
##
     Domain Functional_type Sample_ID USDA_Species_Code LMA_gDW_m2 PLSR_LMA_gDW_m2
## 1
        D02
                  broadleaf
                                 P0001
                                                     JUNI
                                                               72.87
                                                                             96.26243
## 2
        D02
                                 L0001
                                                     JUNI
                                                               72.87
                                                                             90.09453
                  broadleaf
## 3
        D02
                  broadleaf
                                 P0002
                                                     JUNI
                                                               60.77
                                                                             77.16475
                                                     JUNI
                                                               60.77
## 4
        D02
                  broadleaf
                                 L0002
                                                                             60.99039
```

```
PLSR Residuals
##
                         LCI
                                    UCI
                                             LPI
                                                      UPI
         23.3924343 93.95423 99.03625 71.30476 121.2201
## 1
## 2
         17.2245326 88.81329 92.00078 65.21071 114.9784
## 3
         16.3947533 74.79509 79.85715 52.19722 102.1323
## 4
         0.2203913 59.80058 62.29402 36.12678 85.8540
         15.3070857 98.86570 103.97701 76.26586 126.1883
## 5
## 6
         11.2101840 95.50843 99.66865 72.20971 122.0507
rmsep_percrmsep <- spectratrait::percent_rmse(plsr_dataset = LeafLMA.PLSR.dataset,</pre>
                                               inVar = inVar,
                                               residuals = LeafLMA.PLSR.dataset$PLSR Residuals,
                                               range="full")
RMSEP <- rmsep_percrmsep$rmse</pre>
perc_RMSEP <- rmsep_percrmsep$perc_rmse</pre>
r2 <- round(summary(lm(LeafLMA.PLSR.dataset$PLSR LMA gDW m2~
                         LeafLMA.PLSR.dataset[,inVar]))$adj.r.squared,2)
expr <- vector("expression", 3)</pre>
expr[[1]] \leftarrow bquote(R^2==.(r2))
expr[[2]] <- bquote(RMSEP==.(round(RMSEP,2)))</pre>
expr[[3]] <- bquote("%RMSEP"==.(round(perc_RMSEP,2)))</pre>
rng_vals <- c(min(LeafLMA.PLSR.dataset$LPI), max(LeafLMA.PLSR.dataset$UPI))</pre>
par(mfrow=c(1,1), mar=c(4.2,5.3,1,0.4), oma=c(0, 0.1, 0, 0.2))
plotrix::plotCI(LeafLMA.PLSR.dataset$PLSR_LMA_gDW_m2,LeafLMA.PLSR.dataset[,inVar],
                li=LeafLMA.PLSR.dataset$LPI, ui=LeafLMA.PLSR.dataset$UPI, gap=0.009,sfrac=0.000,
                lwd=1.6, xlim=c(rng_vals[1], rng_vals[2]), ylim=c(rng_vals[1], rng_vals[2]),
                err="x", pch=21, col="black", pt.bg=scales::alpha("grey70",0.7), scol="grey80",
                cex=2, xlab=paste0("Predicted ", paste(inVar), " (units)"),
                ylab=paste0("Observed ", paste(inVar), " (units)"),
                cex.axis=1.5.cex.lab=1.8)
abline(0,1,lty=2,lw=2)
plotrix::plotCI(LeafLMA.PLSR.dataset$PLSR_LMA_gDW_m2,LeafLMA.PLSR.dataset[,inVar],
                li=LeafLMA.PLSR.dataset$LCI, ui=LeafLMA.PLSR.dataset$UCI, gap=0.009,sfrac=0.004,
                lwd=1.6, xlim=c(rng_vals[1], rng_vals[2]), ylim=c(rng_vals[1], rng_vals[2]),
                err="x", pch=21, col="black", pt.bg=scales::alpha("grey70",0.7), scol="black",
```

85.92

85.92

101.22709

97.13018

JUNI

JUNT.

P0003

1.0003

broadleaf

broadleaf

## 5

## 6

D02

D02

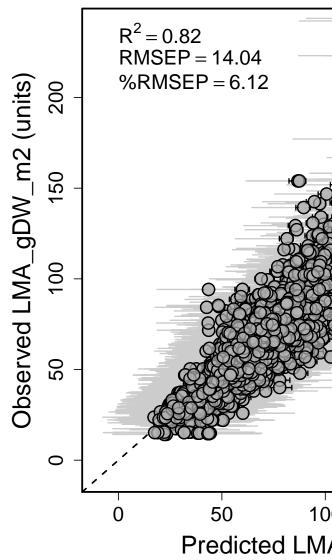
cex=2, xlab=paste0("Predicted ", paste(inVar), " (units)"),

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

cex.axis=1.5,cex.lab=1.8, add=T)

legend("topleft", legend=expr, bty="n", cex=1.5)

box(1wd=2.2)



# Generate PLSR estimated LMA observed vs predicted plot

## 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_PLSR_Estimates.csv"

## [2] "LMA_gDW_m2_PLSR_Validation_Scatterplot.png"
```