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

##

combine

```
list.of.packages <- c("pls", "dplyr", "here", "plotrix", "ggplot2", "gridExtra", "spectratrait")</pre>
invisible(lapply(list.of.packages, library, character.only = TRUE))
## Warning: package 'pls' was built under R version 4.3.1
##
## Attaching package: 'pls'
## The following object is masked from 'package:stats':
##
##
       loadings
## Warning: package 'dplyr' was built under R version 4.3.1
## 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/Library/CloudStorage/OneDrive-NASA/Data/Github/spectratrait
## Warning: package 'plotrix' was built under R version 4.3.1
## Warning: package 'ggplot2' was built under R version 4.3.1
##
## Attaching package: 'gridExtra'
## The following object is masked from 'package:dplyr':
##
```

Setup other functions and options

```
### Setup options
# 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/th/fpt_z3417gn8xgply92pvy6r0000gq/T/Rtmpm6EsT5"

Grab data from EcoSIS

Delimiter: ","

```
print(paste0("Output directory: ",getwd())) # check wd
```

 $\label{lem:ura} \begin{tabular}{ll} URL: $https://ecosis.org/package/fresh-leaf-spectra-to-estimate-lma-over-neon-domains-ineastern-united-states \end{tabular}$

```
## dbl (2152): LMA, 350, 351, 352, 353, 354, 355, 356, 357, 358, 359, 360, 361,...
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
## Download complete!
```

(10): Affiliation, Common Name, Domain, Functional_type, Latin Genus, ...

```
head(dat_raw)
## # A tibble: 6 x 2,162
     Affiliation
                            `Common Name` Domain Functional_type
                                                                    LMA `Latin Genus`
     <chr>
                            <chr>
                                          <chr>
                                                 <chr>>
                                                                  <dbl> <chr>
## 1 University of Wiscon~ black walnut DO2
                                                 broadleaf
                                                                   72.9 Juglans
## 2 University of Wiscon~ black walnut DO2
                                                 broadleaf
                                                                   72.9 Juglans
                                                                   60.8 Juglans
## 3 University of Wiscon~ black walnut DO2
                                                 broadleaf
                                                                   60.8 Juglans
## 4 University of Wiscon~ black walnut DO2
                                                 broadleaf
## 5 University of Wiscon~ black walnut DO2
                                                 broadleaf
                                                                   85.9 Juglans
## 6 University of Wiscon~ black walnut DO2
                                                 broadleaf
                                                                   85.9 Juglans
## # i 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>, ...
## #
names(dat raw)[1:40]
##
    [1] "Affiliation"
                           "Common Name"
                                             "Domain"
                                                                "Functional_type"
   [5] "LMA"
                           "Latin Genus"
                                             "Latin Species"
                                                                "PI"
  [9] "Project"
                                                                "350"
                           "Sample_ID"
                                             "USDA Symbol"
##
## [13] "351"
                           "352"
                                             "353"
                                                                "354"
                           "356"
                                             "357"
                                                                "358"
## [17] "355"
## [21] "359"
                           "360"
                                             "361"
                                                                "362"
                                             "365"
## [25] "363"
                           "364"
                                                                "366"
## [29] "367"
                           "368"
                                             "369"
                                                                "370"
## [33] "371"
                           "372"
                                             "373"
                                                                "374"
## [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
                            `Common Name` Domain Functional_type
##
     Affiliation
                                                                    LMA `Latin Genus`
##
     <chr>>
                            <chr>>
                                          <chr> <chr>
                                                                  <dbl> <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
                                                                   60.8 Juglans
## 4 University of Wiscon~ black walnut DO2
                                                 broadleaf
## 5 University of Wiscon~ black walnut DO2
                                                 broadleaf
                                                                   85.9 Juglans
## 6 University of Wiscon~ black walnut DO2
                                                 broadleaf
                                                                   85.9 Juglans
## # i 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>
            broadleaf
                             P0001
                                       JUNI
                                                                72.9
## 1 D02
## 2 D02
            broadleaf
                             L0001
                                       JUNI
                                                                 72.9
## 3 D02
            broadleaf
                             P0002
                                       JUNI
                                                                 60.8
                             L0002
## 4 DO2
            broadleaf
                                       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
### Create cal/val datasets
## Make a stratified random sampling in the strata USDA_Species_Code and Domain
method <- "dplyr" #base/dplyr
\# base R - a bit slow
# dplyr - much faster
split_data <- spectratrait::create_data_split(dataset=plsr_data,approach=method, split_seed=2356812,
                                                prop=0.8, group_variables=c("USDA_Species_Code", "Domain")
names(split_data)
## [1] "cal data" "val data"
cal.plsr.data <- split_data$cal_data</pre>
head(cal.plsr.data)[1:8]
     Domain Functional_type Sample_ID USDA_Species_Code LMA_gDW_m2
##
                                                                        Wave 500
## 1
        D08
                  broadleaf
                                 L2644
                                                     ACBA
                                                               44.18 0.04170800
## 2
        D08
                  broadleaf
                                 L2646
                                                     ACBA
                                                               41.71 0.05067800
## 3
        D08
                  broadleaf
                                 L2645
                                                     ACBA
                                                               40.66 0.04701700
## 4
        D08
                  broadleaf
                                 P2639
                                                     ACBA
                                                               44.18 0.04125300
## 5
        D03
                  broadleaf
                                                               52.91 0.03895800
                                 P0614
                                                     ACFL
## 6
        D03
                                 L0609
                                                     ACFL
                                                               81.67 0.04186169
                  broadleaf
       Wave_501
                  Wave_502
## 1 0.04208700 0.04283700
## 2 0.05087600 0.05153500
## 3 0.04718200 0.04766500
## 4 0.04150300 0.04247100
## 5 0.03915100 0.03956200
## 6 0.04217802 0.04258768
val.plsr.data <- split_data$val_data</pre>
head(val.plsr.data)[1:8]
      Domain Functional_type Sample_ID USDA_Species_Code LMA_gDW_m2 Wave_500
##
## 3
         D02
                   broadleaf
                                  P0002
                                                      JUNI
                                                                60.77 0.043758
## 12
         D02
                   broadleaf
                                  1.0006
                                                      JUNI
                                                                 42.54 0.044338
```

```
## 13
         D02
                    broadleaf
                                   P0007
                                                       QUVE
                                                                 106.57 0.015643
## 19
         D02
                    broadleaf
                                   P0010
                                                       PRSE
                                                                  78.82 0.033019
                                   P0011
## 21
         D02
                    broadleaf
                                                       PRSE
                                                                  86.09 0.024819
         D02
                    broadleaf
                                   L0014
                                                       PRSE
                                                                  67.11 0.040095
## 28
##
      Wave_501 Wave_502
## 3 0.044171 0.044869
## 12 0.044748 0.045294
## 13 0.015579 0.015431
## 19 0.033102 0.033245
## 21 0.024826 0.025045
## 28 0.040397 0.040864
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 <- ggplot(data = cal.plsr.data,</pre>
                         aes(x = cal.plsr.data[,paste0(inVar)])) +
  geom_histogram(fill=I("grey50"),col=I("black"),alpha=I(.7)) +
  labs(title=paste0("Calibration Histogram for ",inVar), x = paste0(inVar),
       y = "Count")
val_hist_plot <- ggplot(data = val.plsr.data,</pre>
                         aes(x = val.plsr.data[,paste0(inVar)])) +
  geom_histogram(fill=I("grey50"),col=I("black"),alpha=I(.7)) +
  labs(title=paste0("Validation Histogram for ",inVar), x = paste0(inVar),
       y = "Count")
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`.
   Calibration Histogram for LMA_gDW_m2
                                                    Validation Histogram for LMA_gDW_m2
 600 -
 400
                                                  100 -
Count
                                                Count
 200
                                                  50 -
           50
                                                            50
                                    200
                                                                                        200
                     LMA_gDW_m2
                                                                     LMA_gDW_m2
ggsave(filename = file.path(outdir,paste0(inVar,"_Cal_Val_Histograms.png")),
       plot = histograms, device="png", width = 30, height = 12, units = "cm",
```

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
## 1
        D08
                  broadleaf
                                 L2644
                                                     ACBA
                                                                44.18
## 2
        B00
                  broadleaf
                                 1.2646
                                                     ACBA
                                                                41.71
## 3
        D08
                  broadleaf
                                                                40.66
                                 L2645
                                                     ACBA
## 4
        D08
                  broadleaf
                                 P2639
                                                     ACBA
                                                                44.18
## 5
        D03
                  broadleaf
                                                                52.91
                                 P0614
                                                     ACFL
## 6
        D03
                  broadleaf
                                 L0609
                                                     ACFL
                                                                81.67
val_spec <- as.matrix(val.plsr.data[, which(names(val.plsr.data) %in% paste0("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
##
## 3
         D02
                    broadleaf
                                  P0002
                                                       JUNI
                                                                 60.77
         D02
                    broadleaf
                                  L0006
                                                                 42.54
## 12
                                                       JUNI
## 13
         D02
                    broadleaf
                                  P0007
                                                      QUVE
                                                                106.57
## 19
                   broadleaf
         D02
                                  P0010
                                                      PRSE
                                                                 78.82
## 21
         D02
                   broadleaf
                                  P0011
                                                      PRSE
                                                                 86.09
## 28
         D02
                   broadleaf
                                  L0014
                                                      PRSE
                                                                 67.11
```

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")
```

Calibration Validation 9 8 Mean Reflectance Mean Reflectance Min/Max Min/Max 95% CI 95% CI 8 8 Reflectance (%) Reflectance (%) 9 9 4 4 20 20 1000 1500 2000 1000 500 500 1500 2000 Wavelength (nm) Wavelength (nm) dev.copy(png,file.path(outdir,paste0(inVar,'_Cal_Val_Spectra.png')), height=2500, width=4900, res=340) ## quartz_off_screen dev.off(); ## pdf ## 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</pre>
random_seed <- 2356812
seg <- 250
maxComps <- 20
iterations <- 40
prop <- 0.70
if (method=="pls") {
  nComps <- spectratrait::find_optimal_components(dataset=cal.plsr.data, targetVariable=inVar,
                                                   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: 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"
 6e+05
 5e+05
SS 4e+05
 3e+05
 2e+05
 1e+05
                                           Number of Components
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
Fit final model
### Fit final model
segs <- 100
plsr.out <- plsr(as.formula(paste(inVar,"~","Spectra")),scale=FALSE,ncomp=nComps,
                 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)
```

3 comps

5 comps

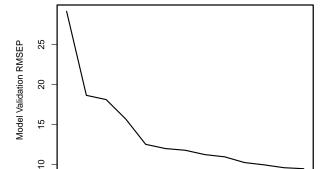
4 comps

2 comps

(Intercept)

1 comps

```
29.174
                      18.644
                                                 15.657
##
                                    18.115
                                                               12.508
                                                                             11.978
##
       6 comps
                     7 comps
                                   8 comps
                                                9 comps
                                                             10 comps
                                                                           11 comps
        11.766
                      11.220
                                    10.941
                                                                9.934
                                                                              9.580
##
                                                 10.226
      12 comps
##
         9.453
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
                                                3 comps
                                                              4 comps
                                                                            5 comps
##
     -0.002151
                    0.590718
                                 0.613614
                                               0.711350
                                                             0.815784
                                                                           0.831077
##
       6 comps
                     7 comps
                                  8 comps
                                                9 comps
                                                             10 comps
                                                                           11 comps
      0.836995
##
                    0.851779
                                 0.859055
                                               0.876874
                                                             0.883790
                                                                           0.891931
##
      12 comps
      0.894776
##
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

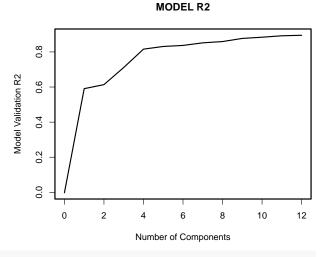
6

Number of Components

8

10

12



par(opar)

0

2

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
        D08
                  broadleaf
                                                                           53.09004
## 1
                                L2644
                                                    ACBA
                                                               44.18
```

```
## 2
        D08
                   broadleaf
                                 L2646
                                                      ACBA
                                                                41.71
                                                                             44.34166
## 3
        D08
                   broadleaf
                                 1.2645
                                                      ACBA
                                                                40.66
                                                                             46.80029
                                 P2639
## 4
        D08
                   broadleaf
                                                      ACBA
                                                                44.18
                                                                             50.07397
        D03
                                                                             64.87644
## 5
                   broadleaf
                                 P0614
                                                      ACFL
                                                                52.91
## 6
        D03
                   broadleaf
                                 L0609
                                                      ACFL
                                                                81.67
                                                                             96.80924
     PLSR CV Predicted PLSR CV Residuals
##
              53.10616
                                 8.926162
## 1
              44.39206
## 2
                                 2.682061
## 3
              46.80108
                                 6.141077
## 4
              50.16964
                                 5.989638
## 5
              65.06744
                                 12.157436
## 6
              96.86793
                                 15.197935
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% "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 PLSR_Predicted
## 3
         D02
                    broadleaf
                                  P0002
                                                       JUNI
                                                                 60.77
                                                                              64.26860
## 12
         D02
                    broadleaf
                                  L0006
                                                       JUNI
                                                                 42.54
                                                                              41.53156
## 13
         D02
                    broadleaf
                                  P0007
                                                       QUVE
                                                                106.57
                                                                              99.94629
## 19
         D02
                    broadleaf
                                                                 78.82
                                  P0010
                                                       PRSE
                                                                              89.09997
## 21
         D02
                    broadleaf
                                  P0011
                                                       PRSE
                                                                 86.09
                                                                              84.86398
         D02
                   broadleaf
## 28
                                  L0014
                                                       PRSE
                                                                 67.11
                                                                              67.88234
##
      PLSR_Residuals
## 3
           3.4986044
## 12
          -1.0084424
## 13
          -6.6237126
## 19
          10.2799665
## 21
          -1.2260190
           0.7723426
## 28
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",
                                            linewidth=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, linewidth=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", linewidth=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, linewidth=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", linewidth=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,
                                    linewidth=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", linewidth=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,
                                     linewidth=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)
```

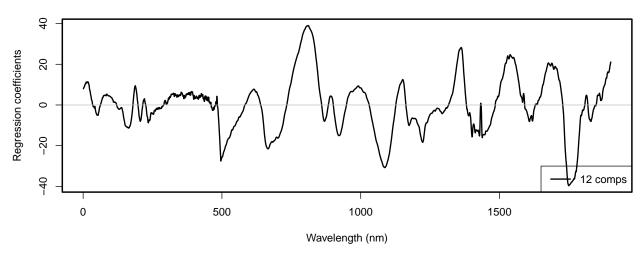
Warning: Removed 22 rows containing missing values or values outside the scale range
(`geom_point()`).

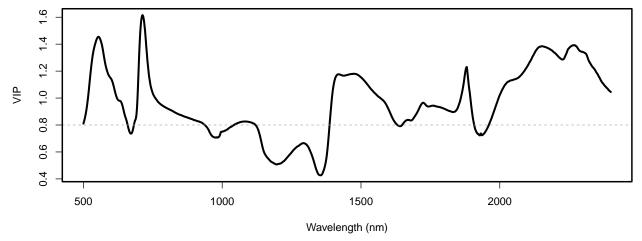
```
## Warning: Removed 8 rows containing missing values or values outside the scale range
## (`geom_point()`).
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
      Calibration: Rsq = 0.88; RMSEP = 10.29
                                                            Validation: Rsq = 0.89; RMSEP = 9.45
                                                        200
Observed LMA_gDW_m2 (units)
                                                         50
              50 100 150 200
Predicted LMA_gDW_m2 (units)
                                                                    50 100 150 Predicted LMA_gDW_m2 (units)
                                                                                                       200
  1500
                                                        200
  1000
count
                                                        100
   500
     0
                    PLSR_CV_Residuals
                                                                           PLSR_Residuals
            -100
                                              50
                                                                   -50
                                                                                                25
ggsave(filename = file.path(outdir,paste0(inVar,"_Cal_Val_scatterplots.png")),
        plot = scatterplots, device="png", width = 32, height = 30, units = "cm",
        dpi = 300)
```

Generate Coefficient and VIP plots

```
vips <- spectratrait::VIP(plsr.out)[nComps,]
par(mfrow=c(2,1))</pre>
```

LMA_gDW_m2





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

pdf ## 2

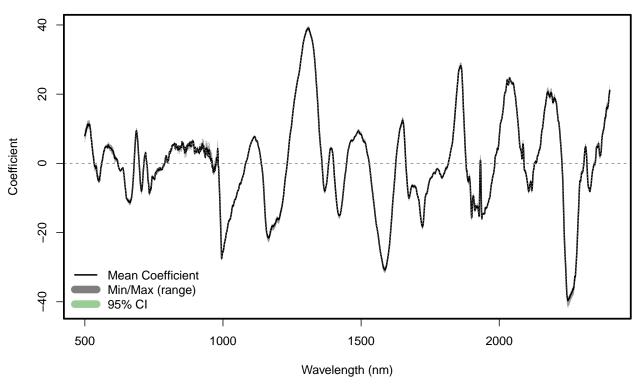
```
par(opar)
```

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$PLSR_Residuals)</pre>
sd tot <- sqrt(sd mean^2+sd res^2)
val.plsr.output$LCI <- Interval_Conf[1,]</pre>
val.plsr.output$UCI <- Interval Conf[2,]</pre>
val.plsr.output$LPI <- val.plsr.output$PLSR_Predicted-1.96*sd_tot
val.plsr.output$UPI <- val.plsr.output$PLSR_Predicted+1.96*sd_tot
head(val.plsr.output)
##
      Domain Functional_type Sample_ID USDA_Species_Code LMA_gDW_m2 PLSR_Predicted
## 3
         D02
                    broadleaf
                                  P0002
                                                      JUNI
                                                                 60.77
                                                                             64.26860
## 12
         D02
                    broadleaf
                                  L0006
                                                      JUNI
                                                                 42.54
                                                                             41.53156
## 13
         D02
                    broadleaf
                                  P0007
                                                      QUVE
                                                                106.57
                                                                             99.94629
         D02
                    broadleaf
                                  P0010
                                                      PRSE
                                                                 78.82
## 19
                                                                             89.09997
## 21
         D02
                    broadleaf
                                  P0011
                                                      PRSE
                                                                 86.09
                                                                             84.86398
                                  L0014
## 28
         D02
                   broadleaf
                                                      PRSE
                                                                 67.11
                                                                             67.88234
##
      PLSR Residuals
                           LCI
                                     UCI
                                               LPI
## 3
           3.4986044 64.14006 64.47482 45.75155 82.78566
## 12
          -1.0084424 41.40129 41.67150 23.01456 60.04855
## 13
          -6.6237126 99.82564 100.08886 81.42930 118.46328
          10.2799665 88.89305 89.27985 70.58206 107.61787
## 19
## 21
          -1.2260190 84.71755 85.02180 66.34672 103.38124
## 28
           0.7723426 67.76659 68.07148 49.36518 86.39950
```

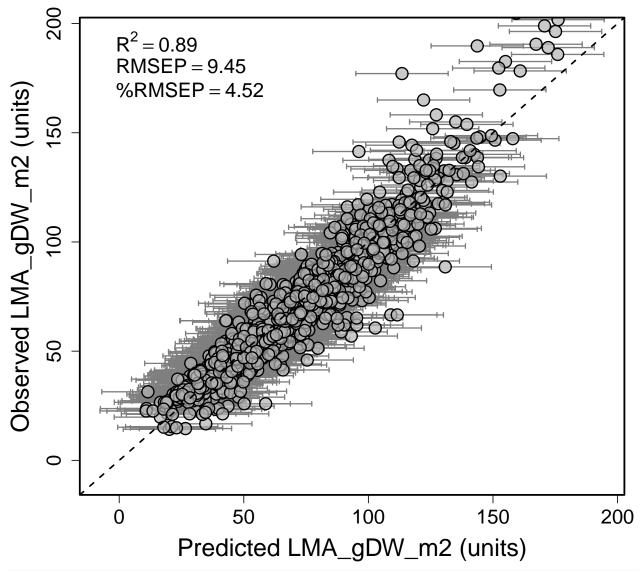
Jackknife coefficient plot

Jackknife regression coefficients



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

Jackknife validation plot



quartz_off_screen

```
##
                   3
dev.off();
## pdf
##
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 68.80319 7.963544 8.372048 8.732315 8.996706
## Seg 1
                 2 68.33966 9.007372 9.404730 9.711349 9.942219
## Seg 2
                3 67.96783 7.809066 8.218030 8.589660 8.855073
## Seg 3
                4 68.56866 7.924350 8.269938 8.602341 8.843661
## Seg 4
## Seg 5
                5 68.06364 7.682055 8.080012 8.388390 8.653853
## Seg 6
                 6 67.73588 7.989043 8.385246 8.743061 8.962190
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/Library/CloudStorage/OneDrive-NASA/Data/Github/spectratrait/v
# 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,
                                          paste0(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
write.csv(vips,file=file.path(outdir,
                               pasteO(inVar,' PLSR VIPs ',
                                      nComps,'comp.csv')))
```

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"
```