Spectra-trait PLSR example using leaf-level spectra and leaf mass per area (LMA) data from more than 40 species grassland species comprising both herbs and graminoids.

Shawn P. Serbin, Julien Lamour, & Jeremiah Anderson

#### 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). In this example, the plants were cultivated in an outdoor setting in the botanical garden of the KIT using 40x40 cm pots with an standardized substrate. The data was measured on a weekly basis (the timestamp is included in the dataset).

## **Getting Started**

#### Installation

```
## Loading required package: usethis
##
## Attaching package: 'remotes'
  The following objects are masked from 'package:devtools':
##
##
       dev_package_deps, install_bioc, install_bitbucket, install_cran,
##
       install_deps, install_dev, install_git, install_github,
##
       install_gitlab, install_local, install_svn, install_url,
##
       install_version, update_packages
## The following object is masked from 'package:usethis':
##
##
       git_credentials
## 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/PLSR_for_plant_trait_prediction
##
## Attaching package: 'gridExtra'
## The following object is masked from 'package:dplyr':
##
## combine
```

### Setup other functions and options

```
### Setup other functions and options
github_dir <- file.path(here::here(), "R_Scripts")</pre>
source_from_gh <- TRUE</pre>
if (source_from_gh) {
  # Source helper functions from GitHub
  print("*** GitHub hash of functions.R file:")
  devtools::source_url("https://raw.githubusercontent.com/TESTgroup-BNL/PLSR_for_plant_trait_prediction
} else {
  functions <- file.path(github_dir, "functions.R")</pre>
  source(functions)
## [1] "*** GitHub hash of functions.R file:"
## SHA-1 hash of file is 2dc61e8f5adc6cd0473eb8f77d3d6c5f2d6a63ae
# 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 <- "SLA_g_cm"</pre>
# What is the source dataset from EcoSIS?
ecosis id <- "3cf6b27e-d80e-4bc7-b214-c95506e46daa"
```

Set working directory (scratch space)

## [1] "Output directory: /private/var/folders/xp/h3k9vf3n2jx181ts786\_yjrn9c2gjq/T/RtmpIdEjB5"

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/PLSR_for_plant_trait_prediction/vignettes"
### Get source dataset from EcoSIS
dat_raw <- get_ecosis_data(ecosis_id = ecosis_id)</pre>
## [1] "**** Downloading Ecosis data ****"
## Downloading data...
## Parsed with column specification:
## cols(
##
     .default = col_double(),
     `growth form` = col_character(),
##
##
     species = col_character(),
##
     timestamp = col_character()
## )
## See spec(...) for full column specifications.
## Download complete!
head(dat_raw)
## # A tibble: 6 x 2,114
     `Anthocyanin co~ `Anthocyanin co~ `Carotenoid con~ `Carotenoid con~
##
                                                                     <dbl>
##
                <dbl>
                                  <dbl>
                                                   <dbl>
## 1
              0.00106
                                 0.997
                                                 0.00799
                                                                      7.49
## 2
              0.00357
                                  1.22
                                                 0.0221
                                                                      7.53
## 3
              0.00252
                                  1.14
                                                 0.0188
                                                                      8.55
## 4
                                                                     11.5
              0.00310
                                  2.26
                                                 0.0158
## 5
              0.00412
                                  1.73
                                                 0.0216
                                                                      9.08
## 6
                                                                      8.66
              0.00397
                                  1.02
                                                 0.0336
     ... with 2,110 more variables: `Chlorophyll concentration (mg/g)` <dbl>,
       `Chlorophyll content ( g/cm )` <dbl>, `LDMC (g/g)` <dbl>, `LFA (mg/cm
       )` <dbl>, `LWC (mg/cm )` <dbl>, `SLA (g/cm )` <dbl>, `growth form` <chr>,
## #
       species <chr>, timestamp <chr>, `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>, '445' <dbl>, '446' <dbl>, '447' <dbl>,
       `448` <dbl>, `449` <dbl>, `450` <dbl>, `451` <dbl>, `452` <dbl>,
       `453` <dbl>, `454` <dbl>, `455` <dbl>, `456` <dbl>, `457` <dbl>,
## #
       `458` <dbl>, `459` <dbl>, `460` <dbl>, `461` <dbl>, `462` <dbl>,
## #
## #
       '463' <dbl>, '464' <dbl>, '465' <dbl>, '466' <dbl>, '467' <dbl>,
## #
       `468` <dbl>, `469` <dbl>, `470` <dbl>, `471` <dbl>, `472` <dbl>,
       `473` <dbl>, `474` <dbl>, `475` <dbl>, `476` <dbl>, `477` <dbl>,
## #
       `478` <dbl>, `479` <dbl>, `480` <dbl>, `481` <dbl>, `482` <dbl>,
## #
```

```
`483` <dbl>, `484` <dbl>, `485` <dbl>, `486` <dbl>, `487` <dbl>,
     `488` <dbl>, `489` <dbl>, `490` <dbl>, ...
names(dat_raw)[1:40]
   [1] "Anthocyanin concentration (mg/g)" "Anthocyanin content ( g/cm )"
   [3] "Carotenoid concentration (mg/g)" "Carotenoid content ( g/cm )"
##
   [5] "Chlorophyll concentration (mg/g)" "Chlorophyll content ( g/cm )"
##
  [7] "LDMC (g/g)"
                                            "LFA (mg/cm )"
## [9] "LWC (mg/cm)"
                                            "SLA (g/cm )"
## [11] "growth form"
                                            "species"
## [13] "timestamp"
                                            "400"
## [15] "401"
                                            "402"
## [17] "403"
                                            "404"
## [19] "405"
                                            "406"
## [21] "407"
                                            "408"
## [23] "409"
                                            "410"
## [25] "411"
                                            "412"
## [27] "413"
                                            "414"
## [29] "415"
                                            "416"
## [31] "417"
                                            "418"
## [33] "419"
                                            "420"
## [35] "421"
                                            "422"
                                            "424"
## [37] "423"
## [39] "425"
                                            "426"
```

### 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])
colnames(Spectra) <- c(paste0("Wave_",wv))</pre>
sample_info <- dat_raw[,names(dat_raw) %notin% seq(350,2500,1)]</pre>
head(sample_info)
## # A tibble: 6 x 13
     `Anthocyanin co~ `Anthocyanin co~ `Carotenoid con~ `Carotenoid con~
##
                <dbl>
                                  <dbl>
                                                    <dbl>
                                                  0.00799
## 1
              0.00106
                                  0.997
                                                                      7.49
## 2
              0.00357
                                  1.22
                                                  0.0221
                                                                      7.53
## 3
              0.00252
                                  1.14
                                                  0.0188
                                                                      8.55
              0.00310
                                  2.26
                                                  0.0158
                                                                      11.5
## 5
              0.00412
                                  1.73
                                                  0.0216
                                                                      9.08
## 6
              0.00397
                                  1.02
                                                  0.0336
                                                                       8.66
## # ... with 9 more variables: `Chlorophyll concentration (mg/g)` <dbl>,
      `Chlorophyll content ( g/cm )` <dbl>, `LDMC (g/g)` <dbl>, `LFA (mg/cm
       )` <dbl>, `LWC (mg/cm )` <dbl>, `SLA (g/cm )` <dbl>, `growth form` <chr>,
## #
       species <chr>, timestamp <chr>
sample_info2 <- sample_info %>%
  select(Plant_Species=species,Growth_Form=`growth form`,timestamp,SLA_g_cm=`SLA (g/cm )`)
head(sample_info2)
```

```
## # A tibble: 6 x 4
##
    Plant_Species
                         Growth_Form timestamp
                                                       SLA_g_cm
##
    <chr>>
                           <chr>
                                       <chr>
                                                          <dbl>
## 1 Calamagrostis epigejos graminoid 5/25/2016 12:20
                                                           107.
## 2 Anthoxanthum odoratum graminoid 5/27/2016 8:40
                                                           293.
## 3 Alopecurus pratensis graminoid 5/27/2016 9:23
                                                           220.
## 4 Festuca ovina
                           graminoid 5/27/2016 9:23
                                                           137.
## 5 Agrostis capillaris
                           graminoid 5/27/2016 9:42
                                                           237.
## 6 Aegopodium podagraria forb
                                       5/25/2016 12:20
                                                           388.
plsr_data <- data.frame(sample_info2,Spectra)</pre>
rm(sample_info, sample_info2, Spectra)
```

Example data cleaning. 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% c(inVar,wv)]),]
# Remove suspect high values
plsr_data <- plsr_data[ plsr_data[,inVar] <= 500, ]</pre>
```

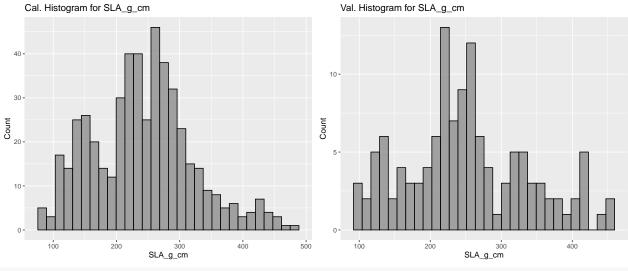
## Create cal/val datasets

```
### Create cal/val datasets
## Make a stratified random sampling in the strata USDA_Species_Code and Domain
method <- "base" #base/dplyr
\# base R - a bit slow
# dplyr - much faster
split_data <- create_data_split(approach=method, split_seed=2356812, prop=0.8,
                                group_variables="Plant_Species")
## Calamagrostis epigejos
                           Cal: 80%
## Anthoxanthum odoratum
                          Cal: 80%
## Alopecurus pratensis
                         Cal: 80%
## Festuca ovina
                  Cal: 78.9473684210526%
## Agrostis capillaris Cal: 82.3529411764706%
## Aegopodium podagraria
                          Cal: 80%
## Arrhenatherum elatius
                           Cal: 82.3529411764706%
## Arctium lappa
                 Cal: 83.3333333333333333
## Urtica dioica
                  Cal: 78.9473684210526%
## Cirsium arvense
                    Cal: 80%
## Geranium pratense
                       Cal: 81.25%
## Geum urbanum
                 Cal: 80%
## Digitalis purpurea
                       Cal: 81.25%
```

```
## Stellaria media
                    Cal: 77.77777777778%
## Trisetum flavescens
                        Cal: 80%
## Trifolium pratense Cal: 80.9523809523809%
## Geranium robertianum
                         Cal: 78.5714285714286%
## Plantago major
                   Cal: 85.7142857142857%
## Nardus stricta
                   Cal: 78.9473684210526%
## Lamium purpureum
                     Cal: 77.777777777778%
## Clinopodium vulgare
                        Cal: 78.5714285714286%
## Poa annua
              Cal: 75%
## Campanula rotundifolia
                         Cal: 78.5714285714286%
## Taraxacum spec.
                    Cal: 80%
                         Cal: 85.7142857142857%
## Digitaria sanguinalis
                   Cal: 82.3529411764706%
## Holcus lanatus
## Lapsana communis
                     Cal: 75%
## Apera spica-venti
                      Cal: 80%
## Alopecurus geniculatus
                           Cal: 75%
## Bromus hordeaceus
                      Cal: 80%
## Phalaris arundinaceae
                         Cal: 81.25%
## Thlaspi arvense Not enough observations
## Origanum vulgare
                     Cal: 77.777777777778%
## Pulicaria dysenterica
                         Cal: 79.166666666667%
## Deschampsia cespitosa
                          Cal: 80%
## Cirsium acaule Cal: 80%
## Brachypodium sylvaticum Cal: 80%
## Centaurium erythraea Cal: 77.777777777778%
## Luzula multiflora
                      Cal: 78.5714285714286%
                        Cal: 78.5714285714286%
## Filipendula ulmaria
## Anthyllis vulneraria
                         Cal: 75%
## Medicago lupulina Cal: 75%
## Scirpus sylvaticus Cal: 77.777777777778%
## Molinia caerulea
                     Cal: 83.3333333333333333
names(split_data)
## [1] "cal_data" "val_data"
cal.plsr.data <- split_data$cal_data</pre>
```

head(cal.plsr.data)[1:8]

```
##
              Plant Species Growth Form
                                              timestamp SLA_g_cm
                                                                   Wave 500
## 1 Calamagrostis epigejos
                              graminoid 5/25/2016 12:20 106.6500 0.09180559
## 2 Anthoxanthum odoratum
                              graminoid 5/27/2016 8:40 293.3565 0.09022668
## 3
      Alopecurus pratensis graminoid 5/27/2016 9:23 220.2703 0.07998340
## 4
              Festuca ovina graminoid 5/27/2016 9:23 137.1220 0.05205080
## 5
                              graminoid 5/27/2016 9:42 237.4237 0.06695127
        Agrostis capillaris
                                   forb 5/25/2016 12:20 388.2384 0.04091566
## 6 Aegopodium podagraria
##
       Wave 501
                  Wave 502
                             Wave 503
## 1 0.09293251 0.09417092 0.09552863
## 2 0.09125158 0.09237300 0.09359694
## 3 0.08109460 0.08231389 0.08365015
## 4 0.05256869 0.05314560 0.05378355
## 5 0.06766205 0.06845248 0.06932220
## 6 0.04169865 0.04257613 0.04355737
val.plsr.data <- split_data$val_data</pre>
head(val.plsr.data)[1:8]
              Plant_Species Growth_Form
                                              timestamp SLA_g_cm
                                                                   Wave_500
## 9
              Urtica dioica
                                   forb 5/25/2016 12:37 284.6788 0.04716736
           Stellaria media
## 15
                                   forb 5/25/2016 13:21 418.4284 0.05694278
## 23 Alopecurus pratensis graminoid 6/1/2016 11:32 218.2117 0.08135086
## 44 Alopecurus pratensis
                             graminoid
                                          6/8/2016 8:37 216.7568 0.10062342
                                          6/8/2016 9:05 231.5292 0.08099724
## 46
        Agrostis capillaris
                              graminoid
## 47 Aegopodium podagraria
                                          6/7/2016 9:05 311.4018 0.03778815
                                   forb
##
        Wave_501
                 Wave_502
                              Wave 503
## 9 0.04781633 0.04854276 0.04935320
## 15 0.05811729 0.05940497 0.06080936
## 23 0.08249180 0.08373915 0.08509719
## 44 0.10190706 0.10330054 0.10480538
## 46 0.08178586 0.08265099 0.08360108
## 47 0.03845043 0.03919155 0.04001581
rm(split_data)
print(paste("Cal observations: ",dim(cal.plsr.data)[1],sep=""))
## [1] "Cal observations: 490"
print(paste("Val observations: ",dim(val.plsr.data)[1],sep=""))
## [1] "Val observations: 124"
cal_hist_plot <- qplot(cal.plsr.data[,paste0(inVar)],geom="histogram",</pre>
                       main = paste0("Cal. Histogram for ",inVar),
                       xlab = paste0(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 = paste0("Val. Histogram for ",inVar),
                       xlab = paste0(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`.
```



## Create calibration and validation PLSR datasets

## 47 Aegopodium podagraria

```
### 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]
##
              Plant_Species Growth_Form
                                               timestamp SLA_g_cm CalVal
## 1 Calamagrostis epigejos
                               graminoid 5/25/2016 12:20 106.6500
                                                                      Cal
## 2
      Anthoxanthum odoratum
                               graminoid 5/27/2016 8:40 293.3565
                                                                      Cal
## 3
       Alopecurus pratensis
                               graminoid 5/27/2016 9:23 220.2703
                                                                      Cal
## 4
              Festuca ovina
                               graminoid 5/27/2016 9:23 137.1220
                                                                      Cal
## 5
        Agrostis capillaris
                               graminoid 5/27/2016 9:42 237.4237
                                                                      Cal
## 6
    Aegopodium podagraria
                                    forb 5/25/2016 12:20 388.2384
                                                                      Cal
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]
              Plant_Species Growth_Form
##
                                               timestamp SLA_g_cm CalVal
## 9
              Urtica dioica
                                    forb 5/25/2016 12:37 284.6788
                                                                      Val
                                    forb 5/25/2016 13:21 418.4284
## 15
            Stellaria media
                                                                      Val
                               graminoid 6/1/2016 11:32 218.2117
                                                                      Val
## 23
       Alopecurus pratensis
## 44
       Alopecurus pratensis
                               graminoid
                                           6/8/2016 8:37 216.7568
                                                                      Val
                                           6/8/2016 9:05 231.5292
## 46
        Agrostis capillaris
                               graminoid
                                                                      Val
```

6/7/2016 9:05 311.4018

Val

forb

## plot cal and val spectra

```
par(mfrow=c(1,2)) # B, L, T, R
f.plot.spec(Z=cal.plsr.data$Spectra,wv=seq(Start.wave,End.wave,1),plot_label="Calibration")
f.plot.spec(Z=val.plsr.data$Spectra,wv=seq(Start.wave,End.wave,1),plot_label="Validation")
                         Calibration
                                                                                  Validation
    100
                                                            100
                                       Mean Reflectance
                                                                                               Mean Reflectance
                                       Min/Max
                                                                                               Min/Max
                                       95% CI
                                                                                               95% CI
    80
                                                            80
                                                        Reflectance (%)
Reflectance (%)
    9
                                                            9
    4
                                                            4
    20
                                                            20
                                                            0
                  1000
                             1500
                                                                           1000
        500
                                        2000
                                                                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 <- "pls" #pls, firstPlateau, firstMin
random_seed <- 2356812
seg <- 100
maxComps <- 18
iterations <- 50
prop <- 0.70
if (method=="pls") {
    # pls package approach - faster but estimates more components...
   nComps <- find_optimal_components(method=method, maxComps=maxComps, seg=seg, random_seed=random_seed)</pre>
```

```
print(paste0("*** Optimal number of components: ", nComps))
} else {
  nComps <- find_optimal_components(dataset=cal.plsr.data, method=method, maxComps=maxComps,
                                     iterations=iterations, seg=seg, prop=prop,
                                     random_seed=random_seed)
}
## [1] "*** Running PLS permutation test ***"
   80
                                                                                  Abs. minimum
                                                                                  Selection
   75
   70
   65
    9
   55
                                5
          0
                                                      10
                                                                            15
                                          Number of components
## [1] "*** Optimal number of components: 10"
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
```

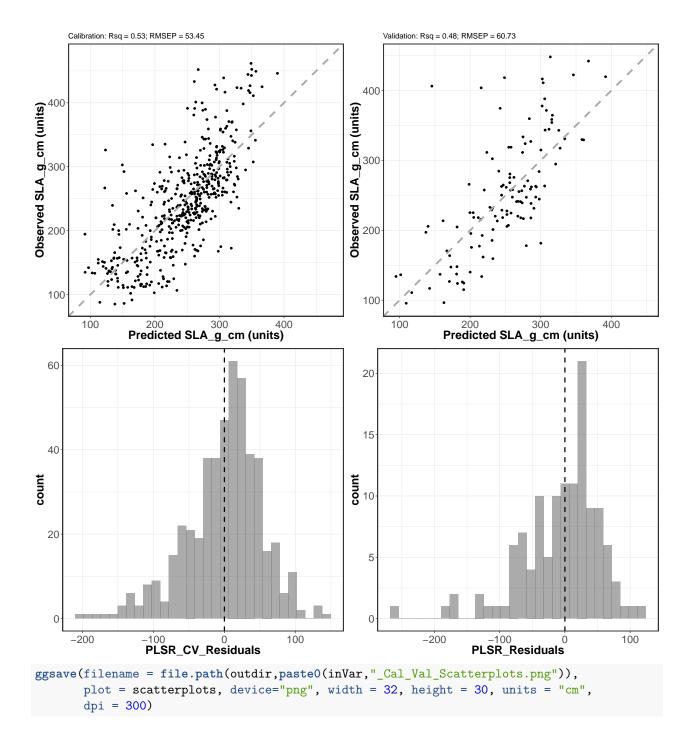
#### Fit final model

```
(Intercept)
                                     2 comps
                                                    3 comps
                      1 comps
                                                                   4 comps
                                                                                  5 comps
                        82.60
                                       81.55
                                                      78.54
                                                                     74.40
                                                                                    69.32
##
          86.06
       6 comps
                      7 comps
                                     8 comps
                                                    9 comps
                                                                  10 comps
##
##
          66.16
                         63.13
                                       61.74
                                                      61.53
                                                                     60.73
plot(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)
R2(plsr.out, newdata = val.plsr.data)
                                                                   4 comps
## (Intercept)
                      1 comps
                                     2 comps
                                                    3 comps
                                                                                  5 comps
##
      -0.01288
                      0.06681
                                     0.09056
                                                    0.15636
                                                                   0.24295
                                                                                  0.34288
##
       6 comps
                      7 comps
                                                                  10 comps
                                     8 comps
                                                    9 comps
       0.40138
                      0.45499
                                     0.47875
                                                    0.48216
                                                                   0.49563
##
plot(R2(plsr.out,estimate=c("test"),newdata = val.plsr.data), main="MODEL R2",
     xlab="Number of Components", ylab="Model Validation R2", lty=1, col="black", cex=1.5, lwd=2)
box(1wd=2.2)
                     MODEL RMSEP
                                                                            MODEL R2
                                                        0.5
    82
                                                        0.4
Model Validation RMSEP
    8
                                                     Model Validation R2
                                                        0.3
   75
                                                        0.2
   2
                                                        0.1
    65
                                                        0.0
    9
        0
               2
                       4
                               6
                                       8
                                              10
                                                            Λ
                                                                    2
                                                                            4
                                                                                    6
                                                                                           8
                                                                                                   10
                    Number of Components
                                                                        Number of Components
dev.copy(png,file.path(outdir,paste0(paste0(inVar,"_Validation_RMSEP_R2_by_Component.png"))),
          height=2800, width=4800, res=340)
## quartz off screen
##
dev.off();
## pdf
par(opar)
```

## PLSR fit observed vs. predicted plot data

```
mutate(PLSR_CV_Residuals = PLSR_CV_Predicted-get(inVar))
head(cal.plsr.output)
##
              Plant_Species Growth_Form
                                               timestamp SLA_g_cm CalVal
## 1 Calamagrostis epigejos
                               graminoid 5/25/2016 12:20 106.6500
                                                                      Cal
     Anthoxanthum odoratum
                               graminoid 5/27/2016 8:40 293.3565
                                                                      Cal
## 3
       Alopecurus pratensis
                               graminoid 5/27/2016 9:23 220.2703
                                                                      Cal
## 4
                               graminoid 5/27/2016 9:23 137.1220
                                                                      Cal
              Festuca ovina
## 5
        Agrostis capillaris
                              graminoid 5/27/2016 9:42 237.4237
                                                                      Cal
## 6
     Aegopodium podagraria
                                    forb 5/25/2016 12:20 388.2384
                                                                      Cal
##
     PLSR_Predicted PLSR_CV_Predicted PLSR_CV_Residuals
## 1
           231.9307
                              234.1193
                                              127.469378
## 2
           237.6749
                              236.7755
                                              -56.581079
## 3
           262.8365
                              263.8336
                                               43.563272
## 4
           126.5863
                                               -8.283722
                              128.8382
## 5
           251.2489
                              251.3030
                                               13.879308
## 6
           277.2292
                              274.2644
                                             -113.974044
cal.R2 <- round(pls::R2(plsr.out)[[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")],
                               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)
##
             Plant Species Growth Form
                                              timestamp SLA_g_cm CalVal
## 1
                                   forb 5/25/2016 12:37 284.6788
             Urtica dioica
                                                                     Val
## 2
           Stellaria media
                                   forb 5/25/2016 13:21 418.4284
                                                                     Val
## 3 Alopecurus pratensis
                             graminoid 6/1/2016 11:32 218.2117
                                                                     Val
      Alopecurus pratensis
                             graminoid
                                         6/8/2016 8:37 216.7568
                                                                     Val
       Agrostis capillaris
                                          6/8/2016 9:05 231.5292
                                                                     Val
## 5
                              graminoid
                                          6/7/2016 9:05 311.4018
## 6 Aegopodium podagraria
                                   forb
                                                                     Val
##
     PLSR Predicted PLSR Residuals
## 1
           240.6023
                        -44.076512
## 2
           248.6923
                       -169.736117
## 3
           211.4638
                         -6.747881
## 4
           275.4544
                         58.697587
## 5
           290,4019
                         58.872672
           274.2311
## 6
                        -37.170622
val.R2 <- round(pls::R2(plsr.out,newdata=val.plsr.data)[[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, y=get(inVar))) +</pre>
  theme_bw() + geom_point() + geom_abline(intercept = 0, slope = 1, color="dark grey",
                                           linetype="dashed", size=1.5) + xlim(rng quant[1],
                                                                                rng_quant[2]) +
  ylim(rng_quant[1], rng_quant[2]) +
  labs(x=paste0("Predicted ", paste(inVar), " (units)"),
       y=paste0("Observed ", paste(inVar), " (units)"),
```

```
title=paste0("Calibration: ", paste0("Rsq = ", cal.R2), "; ", paste0("RMSEP = ",
                                                                             cal.RMSEP))) +
  theme(axis.text=element_text(size=18), legend.position="none",
        axis.title=element text(size=20, face="bold"),
        axis.text.x = element_text(angle = 0, vjust = 0.5),
        panel.border = element_rect(linetype = "solid", fill = NA, size=1.5))
cal resid histogram <- ggplot(cal.plsr.output, aes(x=PLSR CV Residuals)) +</pre>
  geom histogram(alpha=.5, position="identity") +
  geom_vline(xintercept = 0, color="black",
             linetype="dashed", size=1) + theme_bw() +
  theme(axis.text=element text(size=18), legend.position="none",
        axis.title=element_text(size=20, face="bold"),
        axis.text.x = element_text(angle = 0, vjust = 0.5),
        panel.border = element_rect(linetype = "solid", fill = NA, size=1.5))
rng_quant <- quantile(val.plsr.output[,inVar], probs = c(0.001, 0.999))</pre>
val_scatter_plot <- ggplot(val.plsr.output, aes(x=PLSR_Predicted, y=get(inVar))) +</pre>
  theme_bw() + geom_point() + geom_abline(intercept = 0, slope = 1, color="dark grey",
                                          linetype="dashed", size=1.5) + xlim(rng_quant[1],
                                                                               rng_quant[2]) +
  ylim(rng_quant[1], rng_quant[2]) +
  labs(x=paste0("Predicted ", paste(inVar), " (units)"),
       y=paste0("Observed ", paste(inVar), " (units)"),
       title=paste0("Validation: ", paste0("Rsq = ", val.R2), "; ", paste0("RMSEP = ",
                                                                            val.RMSEP))) +
  theme(axis.text=element_text(size=18), legend.position="none",
        axis.title=element_text(size=20, face="bold"),
        axis.text.x = element_text(angle = 0, vjust = 0.5),
        panel.border = element_rect(linetype = "solid", fill = NA, size=1.5))
val_resid_histogram <- ggplot(val.plsr.output, aes(x=PLSR_Residuals)) +</pre>
  geom_histogram(alpha=.5, position="identity") +
  geom_vline(xintercept = 0, color="black",
             linetype="dashed", size=1) + theme_bw() +
  theme(axis.text=element_text(size=18), legend.position="none",
        axis.title=element_text(size=20, face="bold"),
        axis.text.x = element_text(angle = 0, vjust = 0.5),
        panel.border = element_rect(linetype = "solid", fill = NA, size=1.5))
# plot cal/val side-by-side
scatterplots <- grid.arrange(cal_scatter_plot, val_scatter_plot, cal_resid_histogram,</pre>
                             val resid histogram, nrow=2, ncol=2)
## Warning: Removed 7 rows containing missing values (geom_point).
## Warning: Removed 3 rows containing missing values (geom_point).
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



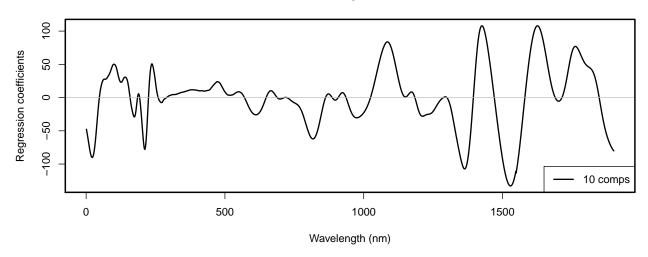
## Generate Coefficient and VIP plots

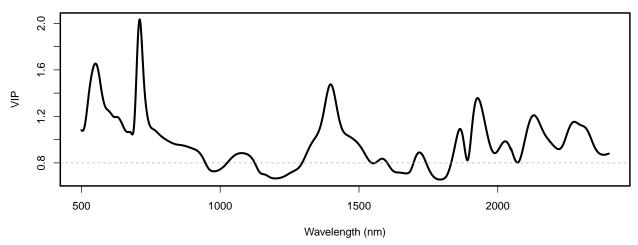
```
vips <- VIP(plsr.out)[nComps,]

par(mfrow=c(2,1))
plot(plsr.out, plottype = "coef",xlab="Wavelength (nm)",
        ylab="Regression coefficients",legendpos = "bottomright",
        ncomp=nComps,lwd=2)
box(lwd=2.2)</pre>
```

```
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)
```

# $SLA_g_cm$





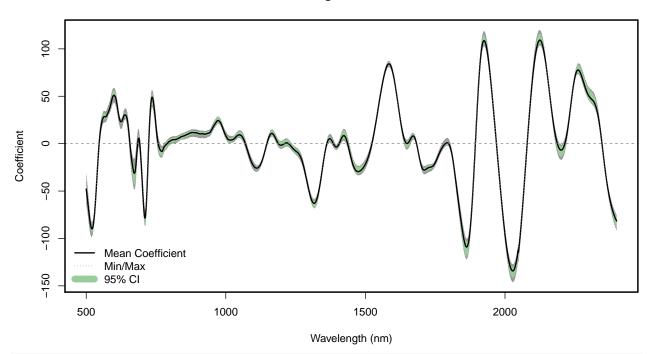
## pdf
## 2
par(opar)

#### Jackknife validation

```
if(grepl("Windows", sessionInfo()$running)){
 pls.options(parallel =NULL)
} else {
  pls.options(parallel = parallel::detectCores()-1)
seg <- 100
jk.plsr.out <- pls::plsr(as.formula(paste(inVar,"~","Spectra")), scale=FALSE,
                         center=TRUE, ncomp=nComps, validation="CV",
                         segments = seg, segment.type="interleaved", trace=FALSE,
                          jackknife=TRUE, data=cal.plsr.data)
pls.options(parallel = NULL)
Jackknife_coef <- f.coef.valid(plsr.out = jk.plsr.out, data_plsr = cal.plsr.data,</pre>
                                ncomp = nComps)
Jackknife_intercept <- Jackknife_coef[1,,,]</pre>
Jackknife coef <- Jackknife coef[2:dim(Jackknife coef)[1],,,]</pre>
interval <-c(0.025, 0.975)
Jackknife_Pred <- val.plsr.data$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, FUN = quantile,</pre>
                       probs=c(interval[1], interval[2]))
sd_mean <- apply(X = Jackknife_Pred, MARGIN = 1, FUN =sd)</pre>
sd_res <- sd(val.plsr.output$PLSR_Residuals)</pre>
sd_tot <- sqrt(sd_mean^2+sd_res^2)</pre>
val.plsr.output$LCI <- Interval_Conf[1,]</pre>
val.plsr.output$UCI <- Interval_Conf[2,]</pre>
val.plsr.output$LPI <- val.plsr.output$PLSR_Predicted-1.96*sd_tot
val.plsr.output$UPI <- val.plsr.output$PLSR_Predicted+1.96*sd_tot
head(val.plsr.output)
##
             Plant_Species Growth_Form
                                              timestamp SLA_g_cm CalVal
## 1
             Urtica dioica
                                   forb 5/25/2016 12:37 284.6788
                                                                     Val
## 2
           Stellaria media
                                   forb 5/25/2016 13:21 418.4284
                                                                     Val
## 3 Alopecurus pratensis graminoid 6/1/2016 11:32 218.2117
                                                                     Val
     Alopecurus pratensis
                             graminoid
                                        6/8/2016 8:37 216.7568
                                                                     Val
       Agrostis capillaris
                                        6/8/2016 9:05 231.5292
## 5
                             graminoid
                                                                     Val
## 6 Aegopodium podagraria
                                  forb
                                        6/7/2016 9:05 311.4018
                                                                     Val
##
     PLSR_Predicted PLSR_Residuals
                                         LCI
                                                  UCI
                                                           LPI
                                                                     UPI
## 1
           240.6023
                        -44.076512 237.5315 250.4949 121.3665 359.8380
## 2
           248.6923
                       -169.736117 246.6740 250.9811 129.6378 367.7468
## 3
           211.4638
                         -6.747881 207.9159 212.8904 92.4012 330.5265
           275.4544
                         58.697587 272.8887 276.9933 156.4053 394.5035
## 4
## 5
           290.4019
                         58.872672 288.2699 291.6463 171.3562 409.4475
## 6
           274.2311
                        -37.170622 272.4991 276.1200 155.1831 393.2792
```

## Jackknife coefficient plot

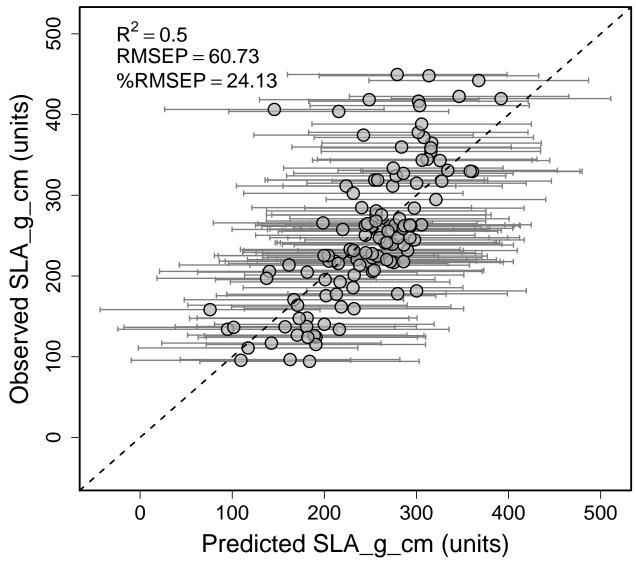
## Jackknife regression coefficients



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

## Jackknife validation plot

```
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=alpha("grey70",0.7), scol="grey50",
    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)
legend("topleft", legend=expr, bty="n", cex=1.5)
box(lwd=2.2)
```



```
## 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
                1 246.6837 -49.80782 -52.32289 -54.88084 -57.63716
## Seg 1
                2 254.8287 -52.24947 -54.31513 -56.41444 -58.71748
## Seg 2
                3 246.2546 -54.91885 -57.12727 -59.35903 -61.78247
## Seg 3
## Seg 4
                4 249.9940 -49.37912 -51.77580 -54.22486 -56.87922
## Seg 5
                5 257.4183 -45.54171 -47.92949 -50.36257 -53.01337
                6 247.2549 -40.72975 -42.81360 -44.93902 -47.28299
## Seg 6
write.csv(out.jk.coefs,file=file.path(outdir,
                                      pasteO(inVar,
                                             '_Jackkife_PLSR_Coefficients.csv')),
          row.names=FALSE)
```

## Create core PLSR outputs

```
print(paste("Output directory: ", getwd()))
## [1] "Output directory: /Users/sserbin/Data/GitHub/PLSR_for_plant_trait_prediction/vignettes"
# 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: "
list.files(outdir)[grep(pattern = inVar, list.files(outdir))]
    [1] "SLA_g_cm_Cal_PLSR_Dataset.csv"
##
   [2] "SLA_g_cm_Cal_Val_Histograms.png"
##
    [3] "SLA_g_cm_Cal_Val_Scatterplots.png"
##
##
    [4] "SLA_g_cm_Cal_Val_Spectra.png"
    [5] "SLA_g_cm_Coefficient_VIP_plot.png"
##
    [6] "SLA_g_cm_Jackkife_PLSR_Coefficients.csv"
##
    [7] "SLA_g_cm_Jackknife_Regression_Coefficients.png"
##
   [8] "SLA_g_cm_Observed_PLSR_CV_Pred_10comp.csv"
##
   [9] "SLA_g_cm_PLSR_Coefficients_10comp.csv"
## [10] "SLA_g_cm_PLSR_Component_Selection.png"
## [11] "SLA_g_cm_PLSR_Validation_Scatterplot.png"
## [12] "SLA_g_cm_PLSR_VIPs_10comp.csv"
## [13] "SLA_g_cm_Val_PLSR_Dataset.csv"
## [14] "SLA_g_cm_Validation_PLSR_Pred_10comp.csv"
## [15] "SLA_g_cm_Validation_RMSEP_R2_by_Component.png"
```