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

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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 specific leaf area (SLA). 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

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/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
# 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"
# Specify output directory, output_dir
# Options:
# tempdir - use a OS-specified temporary directory
# user defined PATH - e.q. "~/scratch/PLSR"
output_dir <- "tempdir"</pre>
```

Set working directory (scratch space)

[1] "Output directory: /private/var/folders/xp/h3k9vf3n2jx181ts786_yjrn9c2gjq/T/RtmpoFulZG"

Grab data from EcoSIS

```
print(paste0("Output directory: ",getwd())) # check wd
## [1] "Output directory: /Users/sserbin/Data/GitHub/PLSR_for_plant_trait_prediction/vignettes"
### Get source dataset from EcoSIS
dat_raw <- spectratrait::get_ecosis_data(ecosis_id = ecosis_id)</pre>
## [1] "**** Downloading Ecosis data ****"
## Downloading data...
##
## cols(
##
    .default = col_double(),
    `growth form` = col_character(),
##
##
    species = col_character(),
##
    timestamp = col_character()
## )
## i Use `spec()` for the full column specifications.
```

```
head(dat_raw)
```

[39] "425"

```
## # A tibble: 6 x 2,114
     `Anthocyanin concen~ `Anthocyanin cont~ `Carotenoid concen~ `Carotenoid conte~
##
                    <dbl>
                                        <dbl>
                                                            <dbl>
                                                                                <dbl>
## 1
                  0.00106
                                        0.997
                                                          0.00799
                                                                                 7.49
## 2
                  0.00357
                                        1.22
                                                          0.0221
                                                                                 7.53
                  0.00252
                                        1.14
                                                          0.0188
                                                                                 8.55
## 4
                  0.00310
                                        2.26
                                                                                11.5
                                                          0.0158
## 5
                  0.00412
                                        1.73
                                                          0.0216
                                                                                 9.08
## 6
                  0.00397
                                        1.02
                                                          0.0336
                                                                                 8.66
## # ... 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"
                                            "400"
## [13] "timestamp"
## [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"
## [37] "423"
                                            "424"
```

"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])</pre>
colnames(Spectra) <- c(paste0("Wave_",wv))</pre>
sample_info <- dat_raw[,names(dat_raw) %notin% seq(350,2500,1)]</pre>
head(sample_info)
## # A tibble: 6 x 13
     `Anthocyanin concen~ `Anthocyanin cont~ `Carotenoid concen~ `Carotenoid conte~
##
                    <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
                  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 <- spectratrait::create_data_split(dataset=plsr_data, 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%
                         Cal: 82.3529411764706%
## Arrhenatherum elatius
## 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: 77.77777777778%
## Clinopodium vulgare
                       Cal: 78.5714285714286%
## Poa annua Cal: 75%
## Campanula rotundifolia
                          Cal: 78.5714285714286%
## Taraxacum spec.
                   Cal: 80%
## Digitaria sanguinalis
                         Cal: 85.7142857142857%
## Holcus lanatus Cal: 82.3529411764706%
## 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: 79.1666666666667%
## Deschampsia cespitosa
                           Cal: 80%
## Cirsium acaule
                   Cal: 80%
## Brachypodium sylvaticum
                             Cal: 80%
## Centaurium erythraea
                         Cal: 77.777777777778%
## Luzula multiflora
                       Cal: 78.5714285714286%
## Filipendula ulmaria
                         Cal: 78.5714285714286%
## Anthyllis vulneraria
                          Cal: 75%
## Medicago lupulina
                       Cal: 75%
## Succisa pratensis
                       Cal: 83.3333333333333333
## Scirpus sylvaticus
                        Cal: 77.777777777778%
## Molinia caerulea
                     Cal: 83.3333333333333333
names(split_data)
## [1] "cal data" "val data"
cal.plsr.data <- split data$cal data
head(cal.plsr.data)[1:8]
              Plant Species Growth Form
                                              timestamp SLA_g_cm
                                                                   Wave 500
## 1 Calamagrostis epigejos
                             graminoid 5/25/2016 12:20 106.6500 0.09180559
## 2 Anthoxanthum odoratum
                             graminoid 5/27/2016 8:40 293.3565 0.09022668
      Alopecurus pratensis
                             graminoid 5/27/2016 9:23 220.2703 0.07998340
## 3
## 4
              Festuca ovina
                             graminoid 5/27/2016 9:23 137.1220 0.05205080
## 5
       Agrostis capillaris
                              graminoid 5/27/2016 9:42 237.4237 0.06695127
    Aegopodium podagraria
                                   forb 5/25/2016 12:20 388.2384 0.04091566
       Wave_501
                 Wave_502
                             Wave_503
## 1 0.09293251 0.09417092 0.09552863
## 2 0.09125158 0.09237300 0.09359694
## 3 0.08109460 0.08231389 0.08365015
## 4 0.05256869 0.05314560 0.05378355
## 5 0.06766205 0.06845248 0.06932220
## 6 0.04169865 0.04257613 0.04355737
val.plsr.data <- split_data$val_data</pre>
head(val.plsr.data)[1:8]
                                              timestamp SLA_g_cm
              Plant_Species Growth_Form
                                                                   Wave_500
## 9
                                   forb 5/25/2016 12:37 284.6788 0.04716736
              Urtica dioica
                                   forb 5/25/2016 13:21 418.4284 0.05694278
## 15
           Stellaria media
                             graminoid 6/1/2016 11:32 218.2117 0.08135086
## 23 Alopecurus pratensis
## 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
                                  forb
                                          6/7/2016 9:05 311.4018 0.03778815
```

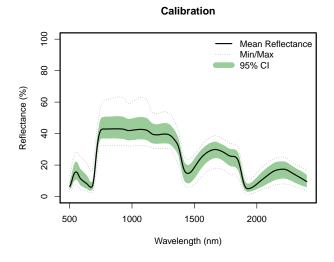
```
##
        Wave 501
                   Wave_502
                               Wave 503
## 9 0.04781633 0.04854276 0.04935320
## 15 0.05811729 0.05940497 0.06080936
## 23 0.08249180 0.08373915 0.08509719
## 44 0.10190706 0.10330054 0.10480538
## 46 0.08178586 0.08265099 0.08360108
## 47 0.03845043 0.03919155 0.04001581
rm(split data)
# Datasets:
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 = pasteO(inVar),ylab = "Count",fill=I("grey50"),
                        col=I("black"),alpha=I(.7))
histograms <- grid.arrange(cal_hist_plot, val_hist_plot, ncol=2)</pre>
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
   Cal. Histogram for SLA_g_cm
                                                  Val. Histogram for SLA_g_cm
 30
                                               Count
      100
               200
                                   400
                                                               200
                     SLA_g_cm
                                                                     SLA_g_cm
ggsave(filename = file.path(outdir,paste0(inVar,"_Cal_Val_Histograms.png")),
       plot = histograms, device="png", width = 30, height = 12, units = "cm",
       dpi = 300)
# output cal/val data
write.csv(cal.plsr.data,file=file.path(outdir,paste0(inVar,'_Cal_PLSR_Dataset.csv')),
          row.names=FALSE)
write.csv(val.plsr.data,file=file.path(outdir,paste0(inVar,'_Val_PLSR_Dataset.csv')),
```

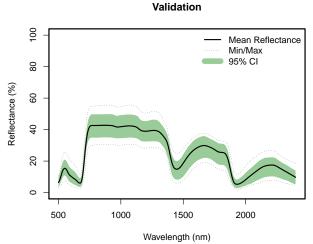
Create calibration and validation PLSR datasets

```
### 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]
##
              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
                                                                      Cal
       Alopecurus pratensis
                               graminoid 5/27/2016 9:23 220.2703
## 3
## 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
                                                                      Val
## 15
            Stellaria media
                                    forb 5/25/2016 13:21 418.4284
                                                                      Val
## 23
       Alopecurus pratensis
                               graminoid 6/1/2016 11:32 218.2117
       Alopecurus pratensis
                                           6/8/2016 8:37 216.7568
                                                                      Val
## 44
                               graminoid
## 46
        Agrostis capillaris
                               graminoid
                                           6/8/2016 9:05 231.5292
                                                                      Val
                                           6/7/2016 9:05 311.4018
                                                                      Val
## 47 Aegopodium podagraria
                                    forb
```

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")
```





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</pre>
random_seed <- 2356812</pre>
seg <- 100
maxComps <- 18
iterations <- 50
prop <- 0.70
if (method=="pls") {
  \# pls package approach - faster but estimates more components....
  nComps <- spectratrait::find_optimal_components(dataset=cal.plsr.data,method=method,</pre>
                                                    maxComps=maxComps, seg=seg,
                                                    random_seed=random_seed)
  print(paste0("*** Optimal number of components: ", nComps))
} else {
  nComps <- spectratrait::find_optimal_components(dataset=cal.plsr.data, method=method,
                                                    maxComps=maxComps,
                                                    iterations=iterations,
                                                    seg=seg, prop=prop,
                                                    random_seed=random_seed)
}
```

[1] "*** Running PLS permutation test ***"

```
Abs. minimum
Selection

22

23

25

26

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)
```

Fit final model

##

dev.off();

pdf ## 2

quartz_off_screen

80

```
segs <- 100
plsr.out <- plsr(as.formula(paste(inVar,"~","Spectra")),scale=FALSE,ncomp=nComps,validation="CV",</pre>
                  segments=segs, segment.type="interleaved",trace=FALSE,data=cal.plsr.data)
fit <- plsr.out$fitted.values[,1,nComps]</pre>
pls.options(parallel = NULL)
# External validation fit stats
par(mfrow=c(1,2)) # B, L, T, R
pls::RMSEP(plsr.out, newdata = val.plsr.data)
## (Intercept)
                     1 comps
                                  2 comps
                                                3 comps
                                                             4 comps
                                                                           5 comps
         86.06
                       82.60
                                    81.55
                                                  78.54
                                                                74.40
                                                                             69.32
##
##
       6 comps
                     7 comps
                                  8 comps
                                                9 comps
                                                            10 comps
         66.16
                       63.13
                                    61.74
##
                                                  61.53
                                                                60.73
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)
```

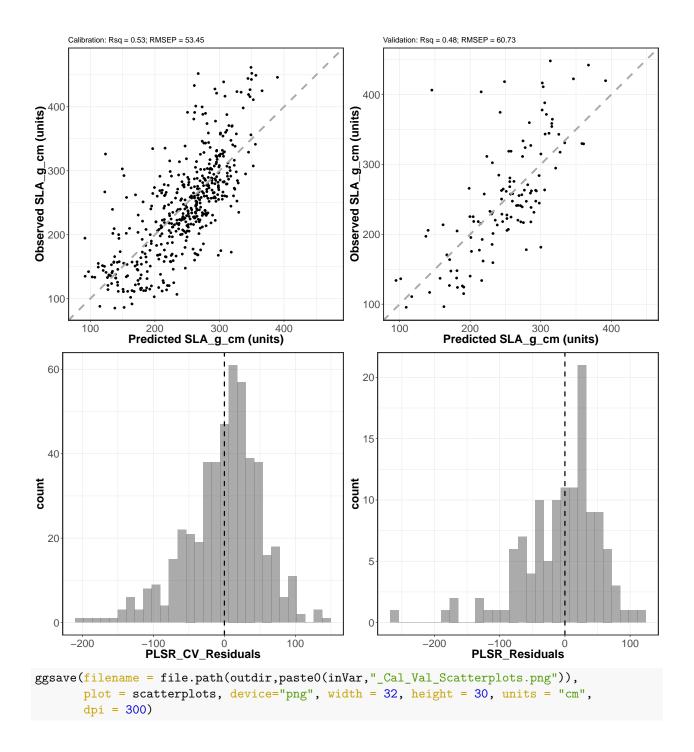
```
pls::R2(plsr.out, newdata = val.plsr.data)
   (Intercept)
                       1 comps
                                       2 comps
                                                      3 comps
                                                                     4 comps
                                                                                    5 comps
##
       -0.01288
                       0.06681
                                       0.09056
                                                      0.15636
                                                                     0.24295
                                                                                    0.34288
##
##
        6 comps
                       7 comps
                                       8 comps
                                                      9 comps
                                                                    10 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
    80
Model Validation RMSEF
                                                      Model Validation R2
                                                          0.3
    75
                                                          0.2
    2
                                                          0.1
    65
                                                          0.0
    9
        0
                2
                                6
                                        8
                                                10
                                                                       2
                                                                                       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
##
                      3
dev.off();
## pdf
##
par(opar)
```

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)
##
              Plant_Species Growth_Form
                                               timestamp SLA_g_cm CalVal
## 1 Calamagrostis epigejos
                                                                     Cal
                              graminoid 5/25/2016 12:20 106.6500
## 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
                                    forb 5/25/2016 12:20 388.2384
                                                                      Cal
    Aegopodium podagraria
     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
                                                43.563272
                              263.8336
## 4
                                                -8.283722
           126.5863
                              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)</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)
##
              Plant_Species Growth_Form
                                                timestamp SLA_g_cm CalVal
## 9
              Urtica dioica
                                    forb 5/25/2016 12:37 284.6788
## 15
            Stellaria media
                                    forb 5/25/2016 13:21 418.4284
                                                                      Val
## 23 Alopecurus pratensis
                              graminoid 6/1/2016 11:32 218.2117
                                                                      Val
                                                                      Val
## 44
       Alopecurus pratensis
                               graminoid
                                           6/8/2016 8:37 216.7568
## 46
        Agrostis capillaris
                               graminoid
                                           6/8/2016 9:05 231.5292
                                                                      Val
## 47 Aegopodium podagraria
                                    forb
                                           6/7/2016 9:05 311.4018
                                                                      Val
      PLSR_Predicted PLSR_Residuals
##
## 9
            240.6023
                         -44.076512
## 15
            248.6923
                        -169.736117
## 23
            211.4638
                          -6.747881
            275.4544
                           58.697587
## 44
## 46
            290.4019
                           58.872672
## 47
            274.2311
                          -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>
```

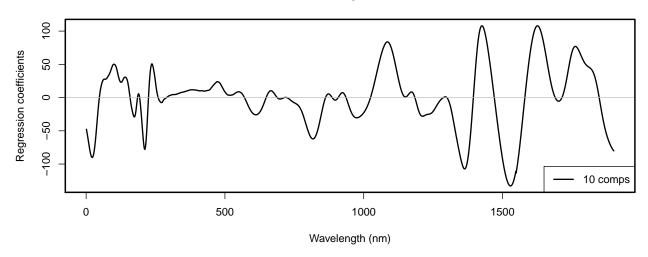
```
geom_histogram(alpha=.5, position="identity") +
  geom_vline(xintercept = 0, color="black",
             linetype="dashed", size=1) + theme_bw() +
  theme(axis.text=element_text(size=18), legend.position="none",
        axis.title=element_text(size=20, face="bold"),
        axis.text.x = element_text(angle = 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`.
```

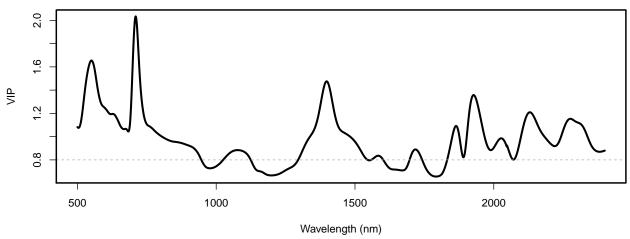


Generate Coefficient and VIP plots

```
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





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

pdf
2
par(opar)

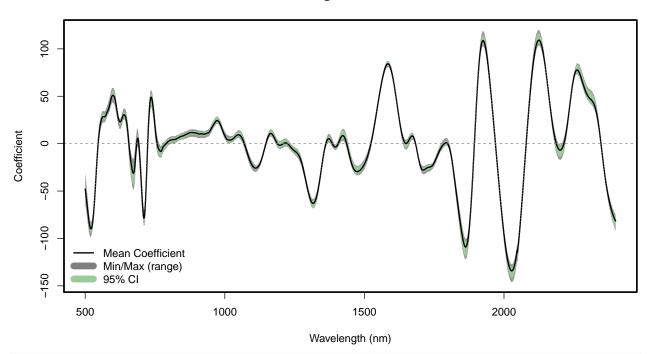
Par (opar)

Jackknife validation

```
if(grepl("Windows", sessionInfo()$running)){
  pls.options(parallel =NULL)
} else {
  pls.options(parallel = parallel::detectCores()-1)
seg <- 100
jk.plsr.out <- pls::plsr(as.formula(paste(inVar,"~","Spectra")), scale=FALSE,
                          center=TRUE, ncomp=nComps, validation="CV",
                          segments = seg, segment.type="interleaved", trace=FALSE,
                          jackknife=TRUE, data=cal.plsr.data)
pls.options(parallel = NULL)
Jackknife_coef <- f.coef.valid(plsr.out = jk.plsr.out, data_plsr = cal.plsr.data,</pre>
                                ncomp = nComps, inVar=inVar)
Jackknife_intercept <- Jackknife_coef[1,,,]</pre>
Jackknife coef <- Jackknife coef[2:dim(Jackknife coef)[1],,,]</pre>
interval <-c(0.025, 0.975)
Jackknife_Pred <- val.plsr.data$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, FUN = quantile,</pre>
                        probs=c(interval[1], interval[2]))
sd_mean <- apply(X = Jackknife_Pred, MARGIN = 1, FUN =sd)</pre>
sd_res <- sd(val.plsr.output$PLSR_Residuals)</pre>
sd_tot <- sqrt(sd_mean^2+sd_res^2)</pre>
val.plsr.output$LCI <- Interval_Conf[1,]</pre>
val.plsr.output$UCI <- Interval_Conf[2,]</pre>
val.plsr.output$LPI <- val.plsr.output$PLSR_Predicted-1.96*sd_tot</pre>
val.plsr.output$UPI <- val.plsr.output$PLSR_Predicted+1.96*sd_tot</pre>
head(val.plsr.output)
##
              Plant_Species Growth_Form
                                               timestamp SLA_g_cm CalVal
## 9
              Urtica dioica
                                    forb 5/25/2016 12:37 284.6788
                                                                       Val
## 15
            Stellaria media
                                    forb 5/25/2016 13:21 418.4284
                                                                       Val
## 23 Alopecurus pratensis graminoid 6/1/2016 11:32 218.2117
                                                                      Val
## 44
       Alopecurus pratensis
                               graminoid
                                          6/8/2016 8:37 216.7568
                                                                      Val
        Agrostis capillaris
                                                                      Val
## 46
                               graminoid
                                          6/8/2016 9:05 231.5292
## 47 Aegopodium podagraria
                                    forb
                                          6/7/2016 9:05 311.4018
                                                                      Val
##
      PLSR_Predicted PLSR_Residuals
                                          LCI
                                                    UCI
                                                             LPI
                                                                      UPI
## 9
            240.6023
                         -44.076512 237.5315 250.4949 121.3665 359.8380
## 15
            248.6923
                        -169.736117 246.6740 250.9811 129.6378 367.7468
## 23
            211.4638
                           -6.747881 207.9159 212.8904 92.4012 330.5265
            275.4544
                           58.697587 272.8887 276.9933 156.4053 394.5035
## 44
## 46
            290.4019
                           58.872672 288.2699 291.6463 171.3562 409.4475
## 47
            274.2311
                         -37.170622 272.4991 276.1200 155.1831 393.2792
```

Jackknife coefficient plot

Jackknife regression coefficients

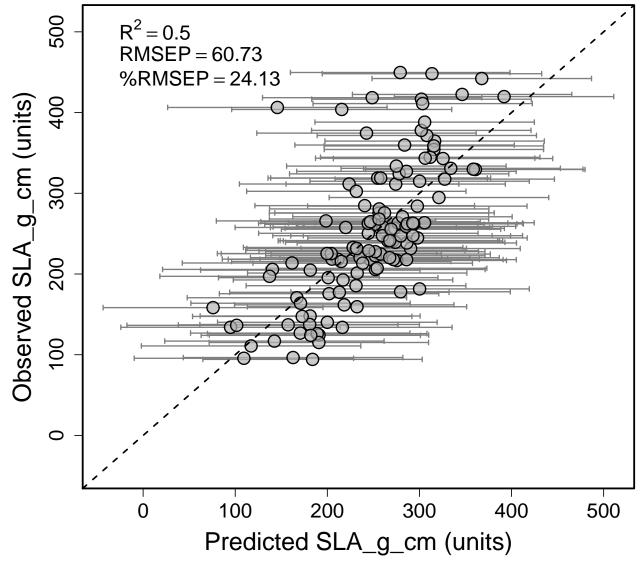


Jackknife validation plot

##

2

```
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="grey50",
    cex=2, xlab=paste0("Predicted ", paste(inVar), " (units)"),
    ylab=paste0("0bserved ", 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,
                                      paste0(inVar,
                                             '_Jackkife_PLSR_Coefficients.csv')),
          row.names=FALSE)
```

Create core PLSR outputs

```
print(paste("Output directory: ", getwd()))
## [1] "Output directory: /Users/sserbin/Data/GitHub/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,
                                         pasteO(inVar,'_Validation_PLSR_Pred_',
                                                 nComps,'comp.csv')),
          row.names=FALSE)
# Model coefficients
coefs <- coef(plsr.out,ncomp=nComps,intercept=TRUE)</pre>
write.csv(coefs,file=file.path(outdir,
                               pasteO(inVar,'_PLSR_Coefficients_',
                                      nComps,'comp.csv')),
          row.names=TRUE)
# PLSR VIP
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] "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"
```