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))
## 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':
##
## combine
```

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 <- "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.g. "~/scratch/PLSR"
output_dir <- "tempdir"</pre>
```

Set working directory (scratch space)

[1] "Output directory: /private/var/folders/th/fpt_z3417gn8xgply92pvy6r0000gq/T/RtmpjZR7Cd"

Grab data from EcoSIS

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!
head(dat_raw)
## # A tibble: 6 x 2,114
     Anthocyanin concentration (mg/~1 Anthocyanin content ~2 Carotenoid concentra~3
##
                                 <dbl>
                                                         <dbl>
                                                                                 <dbl>
## 1
                                                                               0.00799
                               0.00106
                                                         0.997
## 2
                               0.00357
                                                         1.22
                                                                               0.0221
## 3
                               0.00252
                                                         1.14
                                                                               0.0188
## 4
                               0.00310
                                                         2.26
                                                                               0.0158
## 5
                               0.00412
                                                         1.73
                                                                               0.0216
## 6
                               0.00397
                                                         1.02
                                                                               0.0336
## # i abbreviated names: 1: `Anthocyanin concentration (mg/g)`,
       2: `Anthocyanin content ( g/cm )`, 3: `Carotenoid concentration (mg/g)`
## # i 2,111 more variables: `Carotenoid content ( g/cm )` <dbl>,
       `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>, ...
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"
## [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])</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 concentration (mg/~1 Anthocyanin content ~2 Carotenoid concentra~3
                                <dbl>
                                                        <dbl>
##
## 1
                              0.00106
                                                                             0.00799
                                                        0.997
## 2
                              0.00357
                                                        1.22
                                                                             0.0221
## 3
                              0.00252
                                                                             0.0188
                                                        1.14
## 4
                              0.00310
                                                        2.26
                                                                             0.0158
## 5
                              0.00412
                                                        1.73
                                                                             0.0216
## 6
                              0.00397
                                                        1.02
                                                                             0.0336
## # i abbreviated names: 1: `Anthocyanin concentration (mg/g)`,
       2: `Anthocyanin content ( g/cm )`, 3: `Carotenoid concentration (mg/g)`
## # i 10 more variables: `Carotenoid content ( g/cm )` <dbl>,
       `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 )`) %>%
  mutate(SLA_g_cm=as.numeric(SLA_g_cm)) # ensure SLA is numeric
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)
```

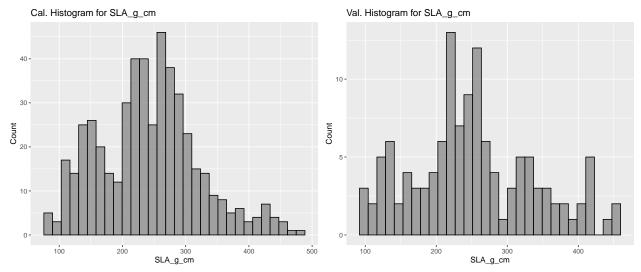
Example data cleaning

```
#### 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

- ## Calamagrostis epigejos Cal: 80%
- ## Anthoxanthum odoratum Cal: 80%
- ## Alopecurus pratensis Cal: 80%
- ## Festuca ovina Cal: 78.947%
- ## Agrostis capillaris Cal: 82.353%
- ## Aegopodium podagraria Cal: 80%
- ## Arrhenatherum elatius Cal: 82.353%
- ## Arctium lappa Cal: 83.333%
- ## Urtica dioica Cal: 78.947%
- ## Cirsium arvense Cal: 80%
- ## Geranium pratense Cal: 81.25%
- ## Geum urbanum Cal: 80%
- ## Digitalis purpurea Cal: 81.25%
- ## Stellaria media Cal: 77.778%
- ## Trisetum flavescens Cal: 80%
- ## Trifolium pratense Cal: 80.952%
- ## Geranium robertianum Cal: 78.571%
- ## Plantago major Cal: 85.714%
- ## Nardus stricta Cal: 78.947%
- ## Lamium purpureum Cal: 77.778%
- ## Clinopodium vulgare Cal: 78.571%
- ## Poa annua Cal: 75%
- ## Campanula rotundifolia Cal: 78.571%
- ## Taraxacum spec. Cal: 80%
- ## Digitaria sanguinalis Cal: 85.714%
- ## Holcus lanatus Cal: 82.353%
- ## 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: 77.778%
- ## Pulicaria dysenterica Cal: 79.167%
- ## Deschampsia cespitosa Cal: 80%
- ## Cirsium acaule Cal: 80%

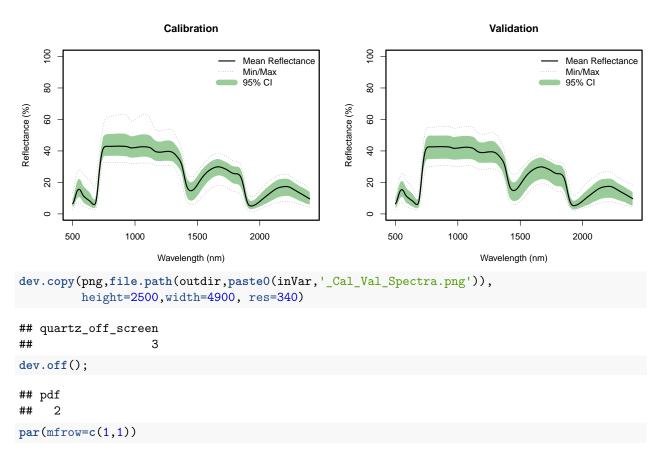
```
## Brachypodium sylvaticum
## Centaurium erythraea
                          Cal: 77.778%
## Luzula multiflora
                       Cal: 78.571%
## Filipendula ulmaria
                         Cal: 78.571%
## Anthyllis vulneraria
                          Cal: 75%
## Medicago lupulina
                       Cal: 75%
## Succisa pratensis
                       Cal: 83.333%
## Scirpus sylvaticus
                        Cal: 77.778%
## Molinia caerulea
                      Cal: 83.333%
names(split_data)
## [1] "cal_data" "val_data"
cal.plsr.data <- split_data$cal_data</pre>
val.plsr.data <- split_data$val_data</pre>
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 = pasteO(inVar),ylab = "Count",fill=I("grey50"),
                       col=I("black"),alpha=I(.7))
## Warning: `qplot()` was deprecated in ggplot2 3.4.0.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
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)
## `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

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
random_seed <- 2356812
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, targetVariable=inVar,
                                                  method=method,
                                                  maxComps=maxComps, seg=seg,
                                                  random_seed=random_seed)
  print(paste0("*** Optimal number of components: ", nComps))
  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 PLS permutation test ***"
    8
                                                                                   Abs. minimum
                                                                                   Selection
   75
   70
    65
    9
   25
          0
                                5
                                                      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
##
Fit final model
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)
                                                                            5 comps
## (Intercept)
                     1 comps
                                   2 comps
                                                3 comps
                                                              4 comps
                                                  78.54
         86.06
                                                                              69.32
##
                       82.60
                                     81.55
                                                                74.40
```

```
##
       6 comps
                      7 comps
                                    8 comps
                                                   9 comps
                                                                10 comps
                                                                    60.73
##
          66.16
                        63.13
                                       61.74
                                                     61.53
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
                                                                                0.34288
##
                                                                 0.24295
##
       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(lwd=2.2)
                     MODEL RMSEP
                                                                          MODEL R2
                                                       0.5
   82
                                                       0.4
Model Validation RMSEP
   80
                                                   Model Validation R2
                                                       0.3
   75
                                                       0.2
   2
                                                       0.1
   65
   9
        0
               2
                       4
                              6
                                      8
                                             10
                                                           0
                                                                   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
#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 PLSR_Predicted
                              graminoid 5/25/2016 12:20 106.6500
                                                                         231.9307
## 1 Calamagrostis epigejos
## 2 Anthoxanthum odoratum
                              graminoid 5/27/2016 8:40 293.3565
                                                                         237.6749
## 3
       Alopecurus pratensis
                              graminoid 5/27/2016 9:23 220.2703
                                                                         262.8365
## 4
              Festuca ovina
                              graminoid 5/27/2016 9:23 137.1220
                                                                         126.5863
## 5
                              graminoid 5/27/2016 9:42 237.4237
                                                                         251.2489
        Agrostis capillaris
    Aegopodium podagraria
                                   forb 5/25/2016 12:20 388.2384
                                                                         277.2292
     PLSR CV Predicted PLSR CV Residuals
##
## 1
              234.1193
                              127.469378
## 2
                              -56.581079
              236.7755
## 3
              263.8336
                               43.563272
## 4
              128.8382
                                -8.283722
## 5
              251.3030
                               13.879308
## 6
              274.2644
                             -113.974044
cal.R2 <- round(pls::R2(plsr.out,intercept=F)[[1]][nComps],2)
cal.RMSEP <- round(sqrt(mean(cal.plsr.output$PLSR_CV_Residuals^2)),2)</pre>
val.plsr.output <- data.frame(val.plsr.data[, which(names(val.plsr.data) %notin% "Spectra")],</pre>
                              PLSR_Predicted=as.vector(predict(plsr.out,
                                                                newdata = val.plsr.data,
                                                                ncomp=nComps, type="response")[,,1]))
val.plsr.output <- val.plsr.output %>%
 mutate(PLSR Residuals = PLSR Predicted-get(inVar))
head(val.plsr.output)
##
              Plant Species Growth Form
                                               timestamp SLA_g_cm PLSR_Predicted
## 9
              Urtica dioica
                                   forb 5/25/2016 12:37 284.6788
                                                                         240.6023
## 15
            Stellaria media
                                   forb 5/25/2016 13:21 418.4284
                                                                         248.6923
## 23 Alopecurus pratensis
                              graminoid 6/1/2016 11:32 218.2117
                                                                         211.4638
## 44 Alopecurus pratensis
                              graminoid
                                           6/8/2016 8:37 216.7568
                                                                         275.4544
        Agrostis capillaris
                              graminoid
                                           6/8/2016 9:05 231.5292
                                                                         290.4019
## 47 Aegopodium podagraria
                                           6/7/2016 9:05 311.4018
                                                                         274.2311
                                   forb
##
      PLSR Residuals
## 9
          -44.076512
## 15
         -169.736117
## 23
           -6.747881
## 44
           58.697587
## 46
           58.872672
          -37.170622
## 47
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, 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))
## Warning: Using `size` aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use `linewidth` instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
## Warning: The `size` argument of `element_rect()` is deprecated as of ggplot2 3.4.0.
## i Please use the `linewidth` argument instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
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)
```

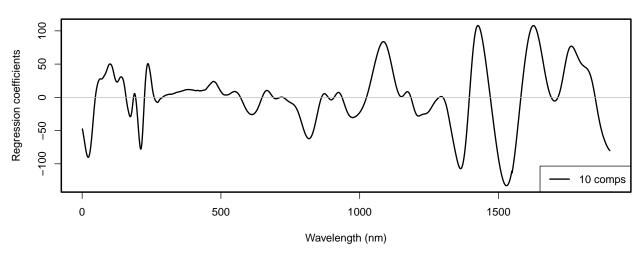
Warning: Removed 7 rows containing missing values or values outside the scale range
(`geom_point()`).

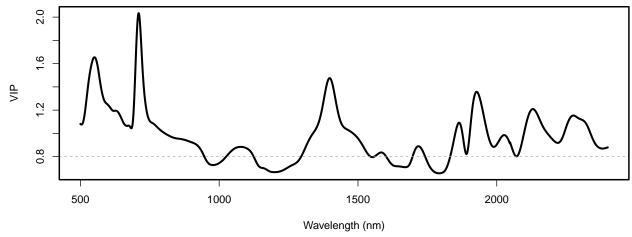
```
## Warning: Removed 3 rows containing missing values or values outside the scale range
## (`geom_point()`).
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
      Calibration: Rsq = 0.54; RMSEP = 53.45
                                                             Validation: Rsq = 0.5; RMSEP = 60.73
                                                          400
Observed SLA_g_cm (units) ^{700}_{500}
                                                       Observed SLA_g_cm (units)
  100
                                                          100
                Predicted SLA_g_cm (units)
                                                                        Predicted SLA_g_cm (units)
        100
                                                               100
  60
                                                          20
                                                          15
  40
                                                       count 10
  20
                                                           5
   0
                   PLSR_CV_Residuals
                                                                             PLSR_Residuals
      -200
                                            100
                                                                     -200
                                                                                                        100
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>
```

SLA_g_cm





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

pdf ## 2

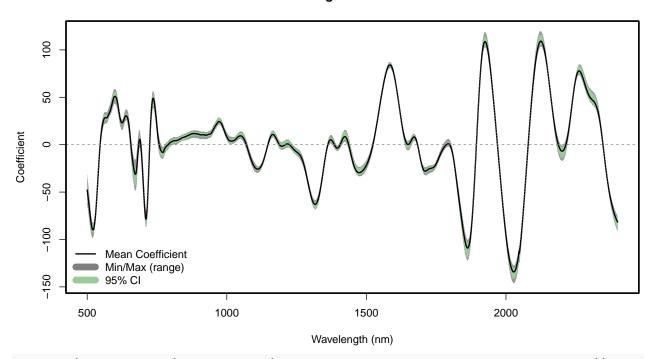
```
par(opar)
```

Jackknife validation

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

Jackknife coefficient plot

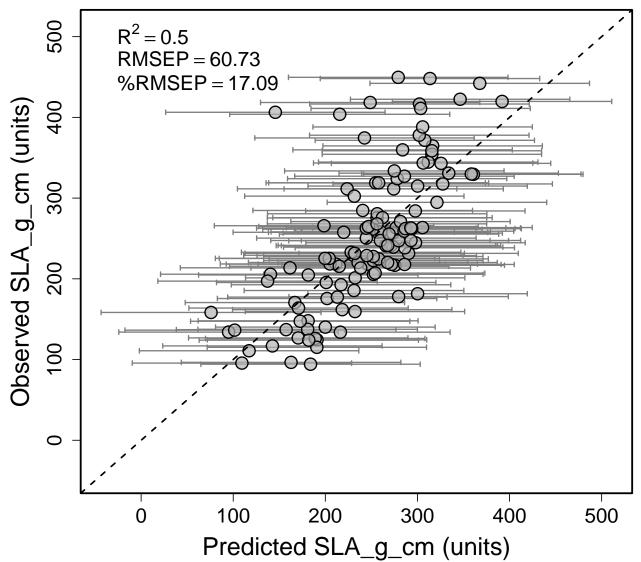
Jackknife regression coefficients



```
## 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
##
## Seg 1
                1 246.6837 -49.80782 -52.32289 -54.88084 -57.63716
## Seg 2
                 2 254.8287 -52.24947 -54.31513 -56.41444 -58.71748
                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
## Seg 6
                 6 247.2549 -40.72975 -42.81360 -44.93902 -47.28299
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,
                              paste0(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"
```