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/spectratrait
##
## 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/tq/tydmhlwn1bdf_0pmpcq70r2c0000gn/T/RtmpNSB86M"

Grab data from EcoSIS

```
print(paste0("Output directory: ",getwd())) # check wd
## [1] "Output directory: /Users/sserbin/Data/Github/spectratrait/vignettes"
### Get source dataset from EcoSIS
dat_raw <- spectratrait::get_ecosis_data(ecosis_id = ecosis_id)</pre>
## [1] "**** Downloading Ecosis data ****"
## Downloading data...
## Rows: 739 Columns: 2114
## -- Column specification -------
## Delimiter: ","
## chr
         (3): growth form, species, timestamp
## dbl (2111): Anthocyanin concentration (mg/g), Anthocyanin content ( g/cm ), ...
## 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
                               0.00106
                                                                              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
                               0.00397
                                                                              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"
## [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"
## [37] "423"
                                            "424"
## [39] "425"
                                            "426"
Create full plsr dataset
### Create plsr dataset
```

```
### Create plsr dataset
Start.wave <- 500
End.wave <- 2400
wv <- seq(Start.wave, End.wave, 1)
Spectra <- as.matrix(dat_raw[,names(dat_raw) %in% wv])
colnames(Spectra) <- c(paste0("Wave_",wv))
sample_info <- dat_raw[,names(dat_raw) %notin% seq(350,2500,1)]
head(sample_info)</pre>
```

```
## # A tibble: 6 x 13
##
    Anthocyanin concentration (mg/~1 Anthocyanin content ~2 Carotenoid concentra~3
##
                                 <dbl>
                                                         <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.
                            graminoid 5/27/2016 9:23
## 3 Alopecurus pratensis
                                                             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

Festuca ovina Cal: 78.947%

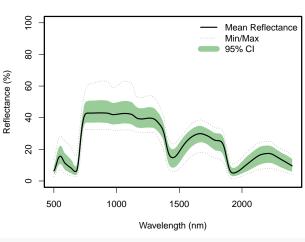
- ## Agrostis capillaris Cal: 82.353%
- ## Aegopodium podagraria Cal: 80%
- ## Arrhenatherum elatius Cal: 82.353%
- ## Arctium lappa Cal: 83.33%
- ## Urtica dioica Cal: 78.947%
- ## Cirsium arvense Cal: 80%
- ## Geranium pratense Cal: 81.25%
- ## Geum urbanum Cal: 80%
- ## Digitalis purpurea Cal: 81.25%
- ## 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 Cal: 80%
- ## Centaurium erythraea Cal: 77.778%
- ## Luzula multiflora Cal: 78.571%
- ## Filipendula ulmaria Cal: 78.571%

```
## Anthyllis vulneraria
                           Cal: 75%
## Medicago lupulina
                        Cal: 75%
                        Cal: 83.333%
## Succisa pratensis
## 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 <- ggplot(data = cal.plsr.data,</pre>
                         aes(x = cal.plsr.data[,paste0(inVar)])) +
  geom_histogram(fill=I("grey50"),col=I("black"),alpha=I(.7)) +
  labs(title=paste0("Calibration Histogram for ",inVar), x = paste0(inVar),
       y = "Count")
val_hist_plot <- ggplot(data = val.plsr.data,</pre>
                         aes(x = val.plsr.data[,paste0(inVar)])) +
  geom_histogram(fill=I("grey50"),col=I("black"),alpha=I(.7)) +
  labs(title=paste0("Validation Histogram for ",inVar), x = paste0(inVar),
       y = "Count")
histograms <- grid.arrange(cal_hist_plot, val_hist_plot, ncol=2)
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
   Calibration Histogram for SLA_g_cm
                                                   Validation Histogram for SLA_g_cm
 40
                                                  10 -
                                                Count
                200
                                    400
                                             500
                                                                200
                                                                                      400
                      SLA_g_cm
                                                                       SLA_g_cm
```

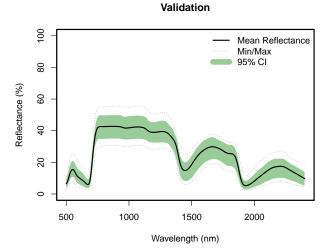
Create calibration and validation PLSR datasets

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")
```

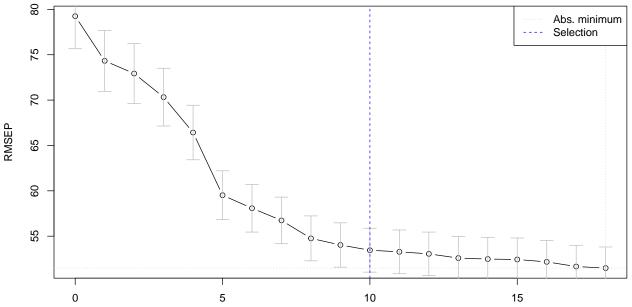


Calibration



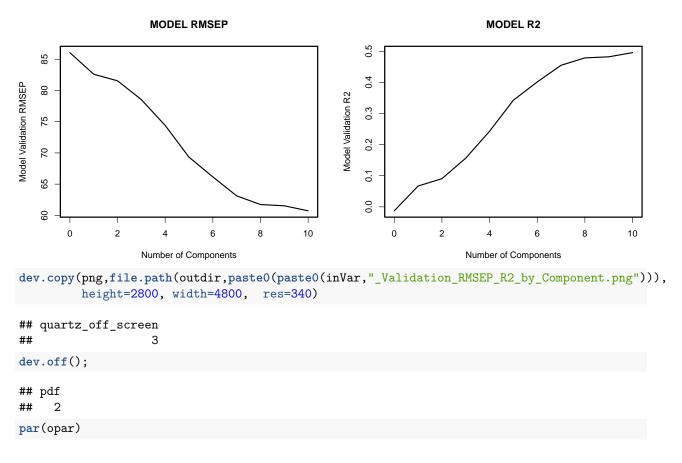
```
## quartz_off_screen
## 3
dev.off();
## pdf
## 2
par(mfrow=c(1,1))
```

```
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</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, targetVariable=inVar,
                                                   method=method,
                                                   maxComps=maxComps, seg=seg,
                                                   random_seed=random_seed)
  print(paste0("*** Optimal number of components: ", nComps))
} else {
  nComps <- spectratrait::find_optimal_components(dataset=cal.plsr.data, targetVariable=inVar,
                                                   method=method,
                                                   maxComps=maxComps,
                                                   iterations=iterations,
                                                   seg=seg, prop=prop,
                                                   random_seed=random_seed)
}
## [1] "*** Identifying optimal number of PLSR components ***"
## [1] "*** Running PLS permutation test ***"
    80
                                                                                 Abs. minimum
                                                                                 Selection
   75
```



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
segs <- 100
plsr.out <- plsr(as.formula(paste(inVar,"~","Spectra")),scale=FALSE,ncomp=nComps,validation="CV",
                 segments=segs, segment.type="interleaved",trace=FALSE,data=cal.plsr.data)
fit <- plsr.out$fitted.values[,1,nComps]</pre>
pls.options(parallel = NULL)
# External validation fit stats
par(mfrow=c(1,2)) # B, L, T, R
pls::RMSEP(plsr.out, newdata = val.plsr.data)
                                                                           5 comps
##
   (Intercept)
                    1 comps
                                  2 comps
                                                3 comps
                                                             4 comps
##
         86.06
                      82.60
                                    81.55
                                                  78.54
                                                                             69.32
                                                               74.40
##
       6 comps
                    7 comps
                                  8 comps
                                                9 comps
                                                            10 comps
                                    61.74
                                                               60.73
         66.16
                       63.13
                                                  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
                                                                           0.34288
##
                                                0.15636
                                                             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(1wd=2.2)
```

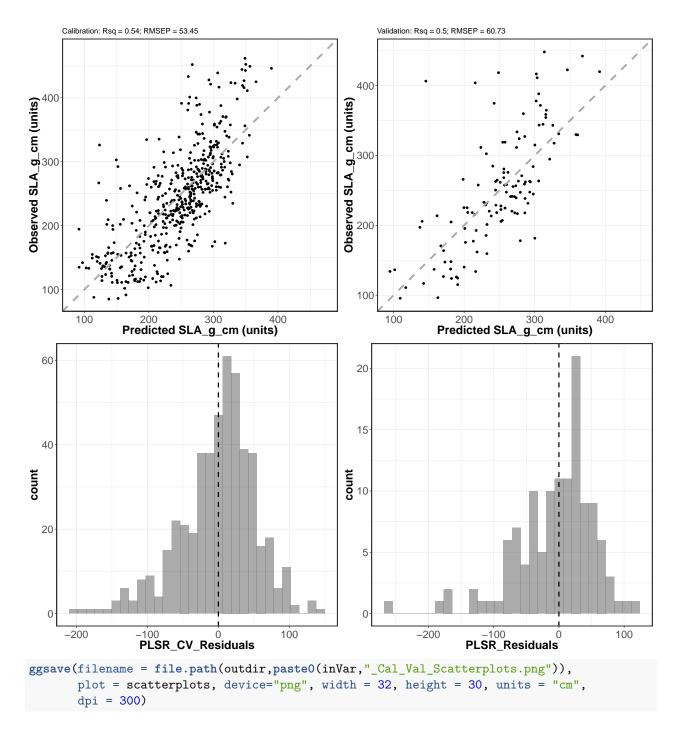


PLSR fit observed vs. predicted plot data

```
#calibration
cal.plsr.output <- data.frame(cal.plsr.data[, which(names(cal.plsr.data) %notin% "Spectra")],</pre>
                               PLSR Predicted=fit,
                               PLSR_CV_Predicted=as.vector(plsr.out$validation$pred[,,nComps]))
cal.plsr.output <- cal.plsr.output %>%
  mutate(PLSR_CV_Residuals = PLSR_CV_Predicted-get(inVar))
head(cal.plsr.output)
##
              Plant_Species Growth_Form
                                                timestamp SLA_g_cm PLSR_Predicted
## 1 Calamagrostis epigejos
                               graminoid 5/25/2016 12:20 106.6500
                                                                          231.9307
## 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
        Agrostis capillaris
                               graminoid 5/27/2016 9:42 237.4237
                                                                          251.2489
##
      Aegopodium podagraria
                                    forb 5/25/2016 12:20 388.2384
                                                                          277.2292
     PLSR_CV_Predicted PLSR_CV_Residuals
##
## 1
              234.1193
                               127.469378
## 2
              236.7755
                               -56.581079
## 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)</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 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
## 46
                                                                        290.4019
## 47 Aegopodium podagraria
                                   forb 6/7/2016 9:05 311.4018
                                                                        274.2311
      PLSR Residuals
## 9
          -44.076512
## 15
         -169.736117
## 23
           -6.747881
## 44
           58.697587
## 46
           58.872672
## 47
          -37.170622
val.R2 <- round(pls::R2(plsr.out,newdata=val.plsr.data,intercept=F)[[1]][nComps],2)
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", linewidth=1.5) +
  xlim(rng_quant[1], rng_quant[2]) +
  ylim(rng_quant[1], rng_quant[2]) +
  labs(x=paste0("Predicted ", paste(inVar), " (units)"),
       y=paste0("Observed ", paste(inVar), " (units)"),
       title=paste0("Calibration: ", paste0("Rsq = ", cal.R2), "; ", paste0("RMSEP = ",
                                                                             cal.RMSEP))) +
  theme(axis.text=element_text(size=18), legend.position="none",
        axis.title=element_text(size=20, face="bold"),
        axis.text.x = element_text(angle = 0, vjust = 0.5),
        panel.border = element_rect(linetype = "solid", fill = NA, linewidth=1.5))
cal_resid_histogram <- ggplot(cal.plsr.output, aes(x=PLSR_CV_Residuals)) +</pre>
  geom_histogram(alpha=.5, position="identity") +
  geom_vline(xintercept = 0, color="black",
             linetype="dashed", linewidth=1) + theme_bw() +
  theme(axis.text=element text(size=18), legend.position="none",
        axis.title=element text(size=20, face="bold"),
        axis.text.x = element_text(angle = 0, vjust = 0.5),
        panel.border = element_rect(linetype = "solid", fill = NA, linewidth=1.5))
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", linewidth=1.5) +
```

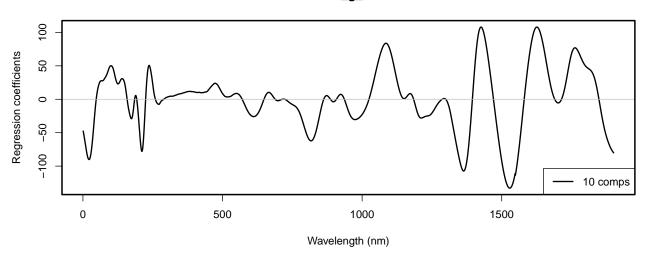
```
xlim(rng_quant[1], rng_quant[2]) +
  ylim(rng_quant[1], rng_quant[2]) +
  labs(x=paste0("Predicted ", paste(inVar), " (units)"),
       y=paste0("Observed ", paste(inVar), " (units)"),
       title=paste0("Validation: ", paste0("Rsq = ", val.R2), "; ", paste0("RMSEP = ",
                                                                            val.RMSEP))) +
  theme(axis.text=element_text(size=18), legend.position="none",
        axis.title=element text(size=20, face="bold"),
        axis.text.x = element_text(angle = 0, vjust = 0.5),
        panel.border = element_rect(linetype = "solid", fill = NA, linewidth=1.5))
val_resid_histogram <- ggplot(val.plsr.output, aes(x=PLSR_Residuals)) +</pre>
  geom_histogram(alpha=.5, position="identity") +
  geom_vline(xintercept = 0, color="black",
             linetype="dashed", linewidth=1) + theme_bw() +
 theme(axis.text=element_text(size=18), legend.position="none",
        axis.title=element_text(size=20, face="bold"),
        axis.text.x = element_text(angle = 0, vjust = 0.5),
        panel.border = element_rect(linetype = "solid", fill = NA, linewidth=1.5))
# plot cal/val side-by-side
scatterplots <- grid.arrange(cal_scatter_plot, val_scatter_plot, cal_resid_histogram,</pre>
                             val_resid_histogram, nrow=2, ncol=2)
## Warning: Removed 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`.
```

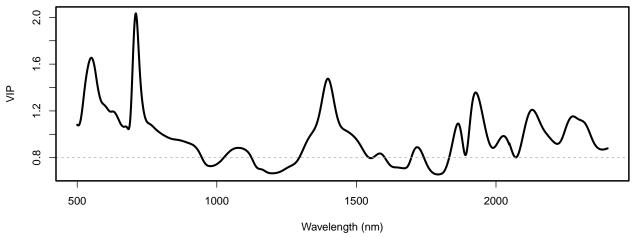


Generate Coefficient and VIP plots

```
lines(seq(Start.wave,End.wave,1),vips,lwd=3)
abline(h=0.8,lty=2,col="dark grey")
box(lwd=2.2)
```

$\mathbf{SLA}_\mathbf{g}_\mathbf{cm}$





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

Jackknife validation

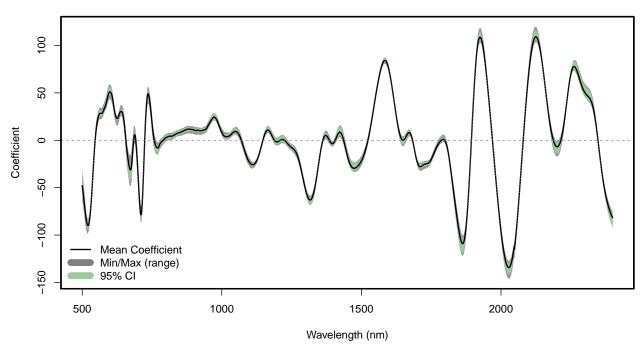
```
if(grepl("Windows", sessionInfo()$running)){
 pls.options(parallel =NULL)
} else {
 pls.options(parallel = parallel::detectCores()-1)
seg <- 100
jk.plsr.out <- pls::plsr(as.formula(paste(inVar, "~", "Spectra")), scale=FALSE,
                          center=TRUE, ncomp=nComps, validation="CV",
                          segments = seg, segment.type="interleaved", trace=FALSE,
                          jackknife=TRUE, data=cal.plsr.data)
pls.options(parallel = NULL)
Jackknife_coef <- f.coef.valid(plsr.out = jk.plsr.out, data_plsr = cal.plsr.data,</pre>
                                ncomp = nComps, inVar=inVar)
Jackknife_intercept <- Jackknife_coef[1,,,]</pre>
Jackknife_coef <- Jackknife_coef[2:dim(Jackknife_coef)[1],,,]</pre>
interval \leftarrow c(0.025, 0.975)
Jackknife_Pred <- val.plsr.data$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 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
## 46
        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
                          LCI
                                    UCI
                                             LPI
## 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
## 46
           58.872672 288.2699 291.6463 171.3562 409.4475
## 47
          -37.170622 272.4991 276.1200 155.1831 393.2792
```

Jackknife coefficient plot

```
spectratrait::f.plot.coef(Z = t(Jackknife_coef), wv = wv,
            plot_label="Jackknife regression coefficients",position = 'bottomleft')
```

```
abline(h=0,lty=2,col="grey50")
box(lwd=2.2)
```

Jackknife regression coefficients

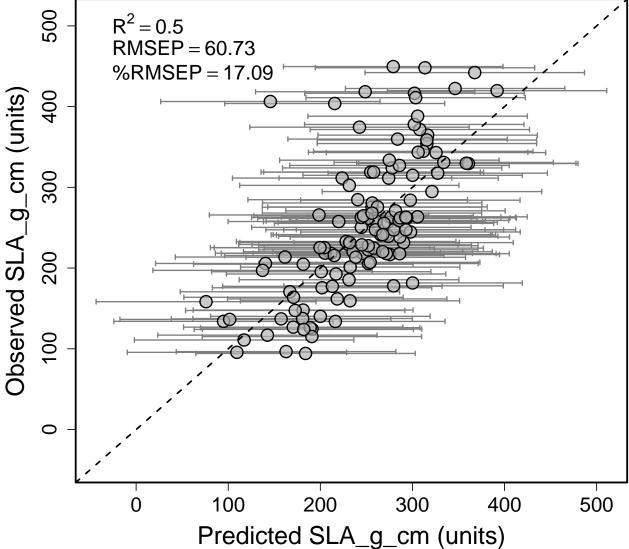


```
## quartz_off_screen
## 3
dev.off();
## pdf
```

Jackknife validation plot

##

```
rmsep_percrmsep <- spectratrait::percent_rmse(plsr_dataset = val.plsr.output,</pre>
                                                inVar = inVar,
                                                residuals = val.plsr.output$PLSR Residuals,
                                                range="full")
RMSEP <- rmsep_percrmsep$rmse
perc_RMSEP <- rmsep_percrmsep$perc_rmse</pre>
r2 <- round(pls::R2(plsr.out, newdata = val.plsr.data, intercept=F)$val[nComps],2)
expr <- vector("expression", 3)</pre>
expr[[1]] <- bquote(R^2==.(r2))
expr[[2]] <- bquote(RMSEP==.(round(RMSEP,2)))</pre>
expr[[3]] <- bquote("%RMSEP"==.(round(perc_RMSEP,2)))</pre>
rng_vals <- c(min(val.plsr.output$LPI), max(val.plsr.output$UPI))</pre>
par(mfrow=c(1,1), mar=c(4.2,5.3,1,0.4), oma=c(0, 0.1, 0, 0.2))
plotrix::plotCI(val.plsr.output$PLSR Predicted,val.plsr.output[,inVar],
       li=val.plsr.output$LPI, ui=val.plsr.output$UPI, gap=0.009,sfrac=0.004,
       lwd=1.6, xlim=c(rng_vals[1], rng_vals[2]), ylim=c(rng_vals[1], rng_vals[2]),
```



```
## 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
## 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
                4 249.9940 -49.37912 -51.77580 -54.22486 -56.87922
## Seg 4
## 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,
                                             '_Jackkife_PLSR_Coefficients.csv')),
          row.names=FALSE)
```

Create core PLSR outputs

```
print(paste("Output directory: ", getwd()))
## [1] "Output directory: /Users/sserbin/Data/Github/spectratrait/vignettes"
# Observed versus predicted
write.csv(cal.plsr.output,file=file.path(outdir,
                                          pasteO(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,
                               paste0(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"
```