Spectra-trait PLSR example using leaf-level spectra and leaf nitrogen content (Narea, g/m2) data from eight different crop species growing in a glasshouse at Brookhaven National Laboratory.

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#### Overview

This is an R Markdown Notebook to illustrate how to load an internal dataset ("ely\_plsr\_data"), choose the "optimal" number of plsr components, and fit a plsr model for leaf nitrogen content (Narea, g/m2)

## **Getting Started**

#### Load libraries

```
list.of.packages <- c("pls","dplyr","here","plotrix","ggplot2","gridExtra","spectratrait")</pre>
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

#### Load internal Ely et al 2019 dataset

```
data("ely plsr data")
head(ely_plsr_data)[,1:8]
##
    Species_Code
                       HEAN3 common sunflower 7.58 15.61210 167.63 36.40 2.103694
## 1
## 2
           HEAN3 common sunflower
                                                        164.68
                                                                  34.65 1.231713
                                      8.33 14.73724
          HEAN3 common sunflower 7.70 15.02495 156.95 35.08 1.764752 CUSA4 garden cucumber 7.40 11.14835 111.52 26.23 1.287963 CUSA4 garden cucumber 7.47 11.60735 123.58 26.71 1.411361
          HEAN3 common sunflower
## 3
## 4
## 5
## 6
            CUSA4 garden cucumber 7.43 8.06035
                                                        114.36 18.40 1.117704
##
   Wave_500
## 1 4.782000
## 2 4.341714
## 3 4.502857
## 4 3.333429
## 5 3.313571
## 6 3.272286
# What is the target variable?
inVar <- "N_g_m2"
```

#### Set working directory (scratch space)

## [1] "/private/var/folders/xp/h3k9vf3n2jx181ts786\_yjrn9c2gjq/T/RtmpKlCiie"

#### Full PLSR dataset

## 10

SOLYL

garden tomato

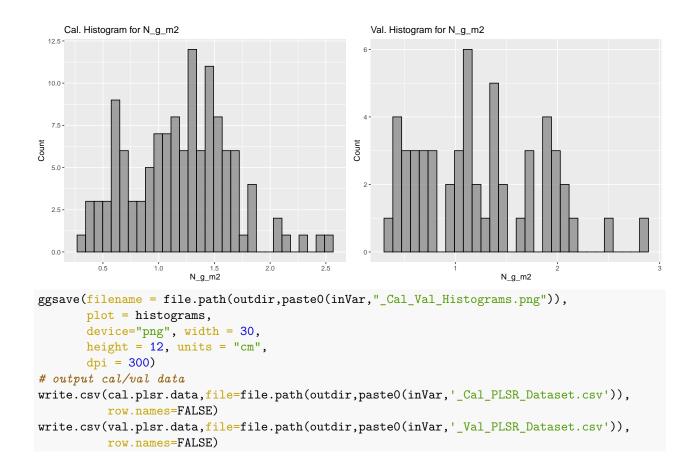
```
Start.wave <- 500
End.wave <- 2400
wv <- seq(Start.wave, End.wave, 1)</pre>
plsr_data <- ely_plsr_data
head(plsr_data)[,1:6]
##
     Species_Code
                       ## 1
                                      7.58 15.61210
            HEAN3 common sunflower
                                                      167.63
                                                                36.40
## 2
           HEAN3 common sunflower
                                      8.33 14.73724
                                                      164.68
                                                                34.65
## 3
           HEAN3 common sunflower
                                      7.70 15.02495
                                                      156.95
                                                                35.08
## 4
           CUSA4 garden cucumber
                                      7.40 11.14835
                                                      111.52
                                                                26.23
## 5
                                      7.47 11.60735
            CUSA4 garden cucumber
                                                      123.58
                                                                26.71
## 6
            CUSA4 garden cucumber
                                      7.43 8.06035
                                                      114.36
                                                                18.40
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=23452135, prop=0.7,
                                             group_variables="Species_Code")
## HEAN3
          Cal: 70%
## CUSA4
          Cal: 68.182%
## CUPE
         Cal: 70.588%
## SOLYL
          Cal: 70%
         Cal: 68.421%
## OCBA
## POPUL
          Cal: 71.429%
          Cal: 70.588%
## GLMA4
## PHVU
         Cal: 66.667%
names(split_data)
## [1] "cal_data" "val_data"
cal.plsr.data <- split_data$cal_data</pre>
head(cal.plsr.data)[1:8]
                        Common_Name C_N_mass C_g_m2 H2O_g_m2 LMA_g_m2 N_g_m2
##
      Species_Code
## 1
            HEAN3 common sunflower
                                       7.58 15.61210
                                                       167.63
                                                                 36.40 2.103694
## 2
                                                       164.68
            HEAN3 common sunflower
                                       8.33 14.73724
                                                                 34.65 1.231713
            CUSA4 garden cucumber
## 4
                                       7.40 11.14835
                                                       111.52
                                                                 26.23 1.287963
## 6
            CUSA4
                   garden cucumber
                                       7.43 8.06035
                                                       114.36
                                                                 18.40 1.117704
## 7
             CUPE
                                       7.20 11.43007
                                                       128.42
                     field pumpkin
                                                                 25.83 1.215333
```

7.89 11.61918

142.23

27.40 1.304110

```
##
      Wave 500
## 1 4.782000
## 2 4.341714
## 4 3.333429
## 6 3.272286
## 7 2.943143
## 10 4.145714
val.plsr.data <- split data$val data</pre>
head(val.plsr.data)[1:8]
##
      Species_Code
                        Common_Name C_N_mass
                                                 C_g_m2 H2O_g_m2 LMA_g_m2
                                                                             N_g_m2
## 3
             HEAN3 common sunflower
                                        7.70 15.024947
                                                          156.95
                                                                    35.08 1.7647515
## 5
             CUSA4 garden cucumber
                                        7.47 11.607347
                                                          123.58
                                                                    26.71 1.4113615
## 8
              CUPE
                      field pumpkin
                                        7.67 12.466238
                                                          124.67
                                                                    29.22 1.1468413
                                        7.64 17.100448
## 9
              CUPE
                      field pumpkin
                                                                    43.39 1.1390174
                                                          142.85
## 13
             SOLYL
                      garden tomato
                                        7.73 7.938866
                                                          129.95
                                                                    17.96 0.9483533
## 15
              OCBA
                        sweet basil
                                        8.13 16.975969
                                                          173.30
                                                                    38.65 1.1246459
##
      Wave_500
## 3 4.502857
## 5 3.313571
## 8 2.868000
## 9 3.338286
## 13 3.960286
## 15 3.744000
rm(split_data)
# Datasets:
print(paste("Cal observations: ",dim(cal.plsr.data)[1],sep=""))
## [1] "Cal observations: 124"
print(paste("Val observations: ",dim(val.plsr.data)[1],sep=""))
## [1] "Val observations: 54"
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))
val_hist_plot <- qplot(val.plsr.data[,paste0(inVar)],geom="histogram",</pre>
                       main = paste0("Val. Histogram for ",inVar),
                       xlab = paste0(inVar),ylab = "Count",fill=I("grey50"),col=I("black"),
                       alpha=I(.7))
histograms <- grid.arrange(cal_hist_plot, val_hist_plot, ncol=2)</pre>
## `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

```
### 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]
##
      Species Code
                         Common_Name C_N_mass
                                                 C_g_m2 H20_g_m2
## 1
             HEAN3 common sunflower
                                          7.58 15.61210
                                                           167.63
## 2
             HEAN3 common sunflower
                                          8.33 14.73724
                                                           164.68
## 4
             CUSA4
                     garden cucumber
                                          7.40 11.14835
                                                           111.52
## 6
             CUSA4
                    garden cucumber
                                         7.43 8.06035
                                                           114.36
## 7
              CUPE
                       field pumpkin
                                          7.20 11.43007
                                                           128.42
## 10
             SOLYL
                       garden tomato
                                         7.89 11.61918
                                                           142.23
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]
##
      Species_Code
                         Common_Name C_N_mass
                                                  C_g_m2 H20_g_m2
## 3
             HEAN3 common sunflower
                                          7.70 15.024947
                                                           156.95
## 5
             CUSA4
                    garden cucumber
                                          7.47 11.607347
                                                            123.58
                                          7.67 12.466238
## 8
              CUPE
                       field pumpkin
                                                           124.67
## 9
              CUPE
                       field pumpkin
                                          7.64 17.100448
                                                           142.85
```

```
## 13 SOLYL garden tomato 7.73 7.938866 129.95
## 15 OCBA sweet basil 8.13 16.975969 173.30
```

#### plot cal and val spectra

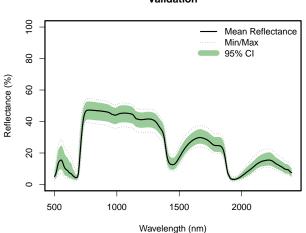
##

```
par(mfrow=c(1,2)) # B, L, T, R
spectratrait::f.plot.spec(Z=cal.plsr.data$Spectra,wv=wv,plot_label="Calibration")
spectratrait::f.plot.spec(Z=val.plsr.data$Spectra,wv=wv,plot_label="Validation")

Calibration

Validation
```

# 



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

# Use 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 <- "firstMin" #firstPlateau, firstMin
random_seed <- 1245565
seg <- 50
maxComps <- 16
iterations <- 80
prop <- 0.70</pre>
```

```
nComps <- spectratrait::find_optimal_comp_by_groups(dataset=cal.plsr.data, targetVariable=inVar,
                                                     method=method, maxComps=maxComps,
                                                     iterations=iterations, prop=prop,
                                                     random_seed=random_seed,
                                                     group_variables="Species_Code")
## [1] "*** Identifying optimal number of PLSR components using stratified resampling by group_variable
## [1] "*** Running permutation test. Please hang tight, this can take awhile ***"
## [1] "Options:"
## [1] "Max Components: 16 Iterations: 80 Data Proportion (percent): 70"
## [1] "*** Providing PRESS and coefficient array output ***"
## No id variables; using all as measure variables
## [1] "*** Optimal number of components based on t.test: 15"
 7.5
 2.5
                                                                       13
                                        Number of Components
dev.copy(png,file.path(outdir,paste0(paste0(inVar,"_PLSR_Component_Selection.png"))),
         height=2800, width=3400, res=340)
## quartz_off_screen
##
dev.off();
## pdf
##
```

### Fit final model

```
fit <- 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)
                                    2 comps
                                                   3 comps
                                                                 4 comps
                                                                                5 comps
##
                      1 comps
                                     0.4162
                                                    0.4037
                                                                                 0.3023
##
        0.5908
                       0.4735
                                                                  0.3347
                                    8 comps
                                                                10 comps
                                                                               11 comps
##
       6 comps
                      7 comps
                                                   9 comps
##
        0.2993
                       0.3081
                                     0.2814
                                                    0.2445
                                                                  0.2276
                                                                                 0.2104
##
      12 comps
                     13 comps
                                   14 comps
                                                  15 comps
        0.1954
                       0.2003
                                     0.1973
                                                    0.2108
##
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
     -0.004079
                     0.355010
                                   0.501632
                                                  0.531088
                                                                0.677620
                                                                              0.737143
##
##
       6 comps
                      7 comps
                                    8 comps
                                                   9 comps
                                                                10 comps
                                                                              11 comps
##
      0.742224
                     0.726835
                                   0.772115
                                                  0.827942
                                                                0.850962
                                                                              0.872685
                                                  15 comps
##
      12 comps
                     13 comps
                                   14 comps
##
      0.890124
                     0.884529
                                   0.887961
                                                  0.872129
plot(pls::R2(plsr.out,estimate=c("test"),newdata = val.plsr.data), main="MODEL R2",
     xlab="Number of Components",ylab="Model Validation R2",lty=1,col="black",cex=1.5,lwd=2)
box(1wd=2.2)
                     MODEL RMSEP
                                                                          MODEL R2
   9.0
                                                       0.8
   0.5
Model Validation RMSEP
                                                   Model Validation R2
                                                       9.0
   0.4
                                                       0.4
   0.3
                                                       0.2
   0.2
                                                       0.0
        0
                    5
                                 10
                                             15
                                                                        5
                                                                                    10
                                                                                                 15
                   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
     2
```

```
par(opar)
```

#### PLSR fit observed vs. predicted plot data

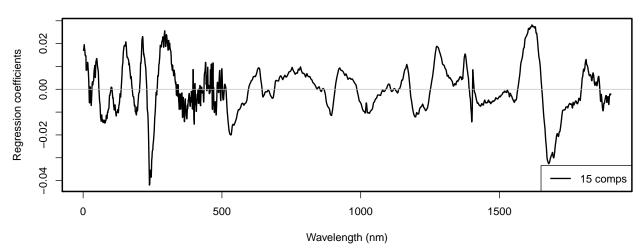
```
#calibration
cal.plsr.output <- data.frame(cal.plsr.data[, which(names(cal.plsr.data) %notin% "Spectra")],</pre>
                               PLSR_Predicted=fit,
                               PLSR_CV_Predicted=as.vector(plsr.out$validation$pred[,,nComps]))
cal.plsr.output <- cal.plsr.output %>%
  mutate(PLSR_CV_Residuals = PLSR_CV_Predicted-get(inVar))
head(cal.plsr.output)
      Species_Code
                                                C_g_m2 H2O_g_m2 LMA_g_m2
##
                         Common_Name C_N_mass
                                                                             N_g_m2
## 1
             HEAN3 common sunflower
                                         7.58 15.61210
                                                          167.63
                                                                     36.40 2.103694
## 2
                                                          164.68
             HEAN3 common sunflower
                                         8.33 14.73724
                                                                     34.65 1.231713
## 4
                    garden cucumber
                                         7.40 11.14835
                                                          111.52
                                                                     26.23 1.287963
             CUSA4
## 6
             CUSA4
                    garden cucumber
                                         7.43 8.06035
                                                          114.36
                                                                     18.40 1.117704
## 7
              CUPE
                       field pumpkin
                                         7.20 11.43007
                                                          128.42
                                                                     25.83 1.215333
## 10
             SOLYL
                       garden tomato
                                         7.89 11.61918
                                                          142.23
                                                                     27.40 1.304110
##
      PLSR_Predicted PLSR_CV_Predicted PLSR_CV_Residuals
## 1
            1.836047
                               1.714086
                                               -0.38960842
## 2
            1.530813
                               1.685388
                                                0.45367526
                                               -0.02512724
## 4
            1.254794
                               1.262835
## 6
            1.127053
                               1.129340
                                                0.01163542
## 7
            1.196259
                               1.188471
                                               -0.02686200
                               1.281683
                                              -0.02242624
## 10
            1.276380
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)
      Species_Code
##
                         Common_Name C_N_mass
                                                  C_g_m2 H20_g_m2 LMA_g_m2
                                                                               N_g_m2
## 3
             HEAN3 common sunflower
                                         7.70 15.024947
                                                           156.95
                                                                      35.08 1.7647515
## 5
             CUSA4
                    garden cucumber
                                         7.47 11.607347
                                                           123.58
                                                                      26.71 1.4113615
## 8
              CUPE
                       field pumpkin
                                         7.67 12.466238
                                                           124.67
                                                                      29.22 1.1468413
## 9
              CUPE
                       field pumpkin
                                         7.64 17.100448
                                                           142.85
                                                                      43.39 1.1390174
## 13
             SOLYL
                       garden tomato
                                         7.73 7.938866
                                                           129.95
                                                                      17.96 0.9483533
              OCBA
                         sweet basil
                                         8.13 16.975969
                                                           173.30
                                                                      38.65 1.1246459
## 15
      PLSR_Predicted PLSR_Residuals
##
## 3
           1.7624701
                        -0.002281391
## 5
           1.2947218
                        -0.116639722
## 8
           0.9934199
                        -0.153421396
## 9
           1.1345273
                        -0.004490078
## 13
                       -0.205067758
           0.7432855
           1.1613789
                         0.036733007
## 15
```

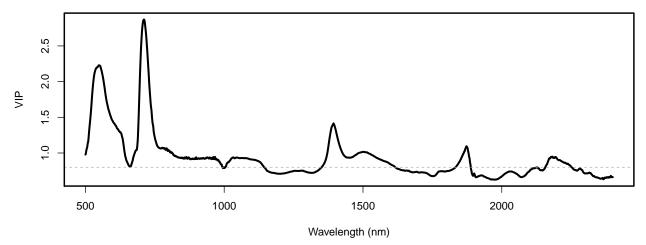
```
val.R2 <- round(pls::R2(plsr.out,newdata=val.plsr.data,intercept=F)[[1]][nComps],2)</pre>
val.RMSEP <- round(sqrt(mean(val.plsr.output$PLSR_Residuals^2)),2)</pre>
rng_quant <- quantile(cal.plsr.output[,inVar], probs = c(0.001, 0.999))</pre>
cal_scatter_plot <- ggplot(cal.plsr.output, aes(x=PLSR_CV_Predicted, 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))
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,
```

```
val_resid_histogram, nrow=2,ncol=2)
## Warning: Removed 5 rows containing missing values (geom_point).
## Warning: Removed 4 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`.
                                                           Validation: Rsq = 0.87; RMSEP = 0.21
     Calibration: Rsq = 0.85; RMSEP = 0.18
  2.5
Observed N_g_m2 (units) 1.0
                                                       Observed N_g_m2 (units)
  0.5
                  1.0 1.5 2.0 Predicted N_g_m2 (units)
                                                  2.5
          0.5
                                                                        Predicted N_g_m2 (units)
  9
count 6
                                                        count
  3
                -0.25 0.00 PLSR_CV_Residuals
                                                                            PLSR_Residuals
     -0.50
                                        0.25
                                                    0.50
                                                                     -0.25
                                                                                                 0.25
ggsave(filename = file.path(outdir,paste0(inVar,"_Cal_Val_Scatterplots.png")),
         plot = scatterplots, device="png",
         width = 32,
         height = 30, units = "cm",
         dpi = 300)
```

# Generate Coefficient and VIP plots

# $N_g_m2$





```
## quartz_off_screen
## 3
```

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

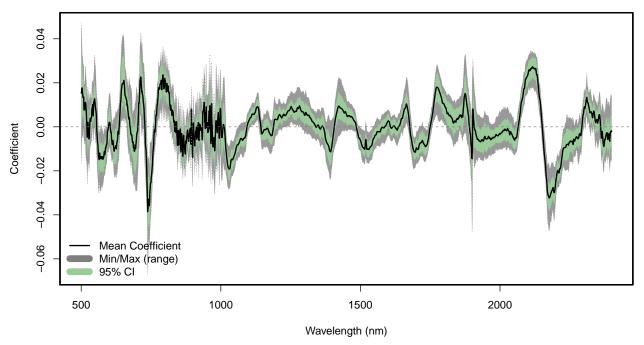
# Bootstrap validation

```
if(grepl("Windows", sessionInfo()$running)){
  pls.options(parallel =NULL)
} else {
  pls.options(parallel = parallel::detectCores()-1)
### PLSR bootstrap permutation uncertainty analysis
iterations <- 500 # how many permutation iterations to run
prop <- 0.70
                      # fraction of training data to keep for each iteration
plsr_permutation <- spectratrait::pls_permutation_by_groups(dataset=cal.plsr.data,</pre>
                                                              targetVariable=inVar,
                                                              maxComps=nComps,
                                                              iterations=iterations,
                                                              prop=prop, group_variables="Species_Code",
                                                              verbose=FALSE)
## [1] "*** Running permutation test. Please hang tight, this can take awhile ***"
## [1] "Options:"
## [1] "Max Components: 15 Iterations: 500 Data Proportion (percent): 70"
## [1] "*** Providing PRESS and coefficient array output ***"
bootstrap_intercept <- plsr_permutation$coef_array[1,,nComps]</pre>
bootstrap_coef <- plsr_permutation$coef_array[2:length(plsr_permutation$coef_array[,1,nComps]),
                                                ,nComps]
rm(plsr_permutation)
# apply coefficients to left-out validation data
interval <-c(0.025, 0.975)
Bootstrap_Pred <- val.plsr.data$Spectra %*% bootstrap_coef +
  matrix(rep(bootstrap_intercept, length(val.plsr.data[,inVar])), byrow=TRUE,
         ncol=length(bootstrap_intercept))
Interval_Conf <- apply(X = Bootstrap_Pred, MARGIN = 1, FUN = quantile,</pre>
                        probs=c(interval[1], interval[2]))
sd_mean <- apply(X = Bootstrap_Pred, MARGIN = 1, FUN = sd)</pre>
sd_res <- sd(val.plsr.output$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
head(val.plsr.output)
##
      Species_Code
                         Common_Name C_N_mass
                                               C_g_m2 H20_g_m2 LMA_g_m2
                                                                              N_g_m2
## 3
             HEAN3 common sunflower
                                         7.70 15.024947 156.95
                                                                     35.08 1.7647515
```

```
SOLYL
                                                                    17.96 0.9483533
## 13
                      garden tomato
                                        7.73 7.938866
                                                          129.95
## 15
              OCBA
                        sweet basil
                                         8.13 16.975969
                                                          173.30
                                                                     38.65 1.1246459
##
      PLSR Predicted PLSR Residuals
                                           LCI
                                                     UCI
                                                               LPI
                                                                        UPI
## 3
           1.7624701
                       -0.002281391 1.5710330 1.9443661 1.3151243 2.209816
## 5
           1.2947218
                       -0.116639722 1.2019841 1.4531979 0.8688563 1.720587
## 8
           0.9934199
                      -0.153421396  0.8544582  1.1646561  0.5564158  1.430424
## 9
           1.1345273
                       -0.004490078 0.9954061 1.2824287 0.7007745 1.568280
                       -0.205067758 0.5836738 0.9094675 0.3042086 1.182362
           0.7432855
## 13
## 15
           1.1613789
                        0.036733007 1.0021191 1.2849671 0.7291004 1.593657
```

# Jackknife coefficient plot

# **Bootstrap regression coefficients**

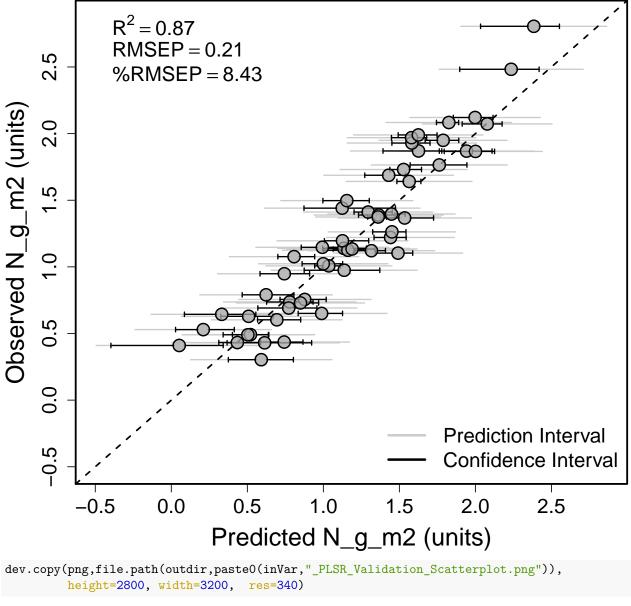


```
## quartz_off_screen
## 3
dev.off();
```

## pdf ## 2

#### Bootstrap 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]] \leftarrow 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$UPI, gap=0.009,sfrac=0.000,
       lwd=1.6, xlim=c(rng_vals[1], rng_vals[2]), ylim=c(rng_vals[1], rng_vals[2]),
       err="x", pch=21, col="black", pt.bg=scales::alpha("grey70",0.7), scol="grey80",
       cex=2, xlab=paste0("Predicted ", paste(inVar), " (units)"),
       ylab=paste0("Observed ", paste(inVar), " (units)"),
       cex.axis=1.5,cex.lab=1.8)
abline(0,1,lty=2,lw=2)
plotrix::plotCI(val.plsr.output$PLSR_Predicted,val.plsr.output[,inVar],
       li=val.plsr.output$LCI, ui=val.plsr.output$UCI, 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]),
       err="x", pch=21, col="black", pt.bg=scales::alpha("grey70",0.7), scol="black",
       cex=2, xlab=paste0("Predicted ", paste(inVar), " (units)"),
       ylab=paste0("Observed ", paste(inVar), " (units)"),
       cex.axis=1.5,cex.lab=1.8, add=T)
legend("topleft", legend=expr, bty="n", cex=1.5)
legend("bottomright", legend=c("Prediction Interval", "Confidence Interval"),
       lty=c(1,1), col = c("grey80","black"), lwd=3, bty="n", cex=1.5)
box(1wd=2.2)
```



```
height=2800, width=3200, res=340)

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

## Output bootstrap results

```
Iteration Intercept
                           Wave 500
                                     Wave 501
                                                Wave 502 Wave 503
           1 0.4731951 0.0236618987 0.021719096 0.023063691 0.02187741
## 1
## 2
           2 0.5415203 -0.0007012397 0.001892634 0.008241293 0.01105366
## 3
           3 0.6512533 0.0123054098 0.013428257 0.015824665 0.01772586
           ## 4
           5 0.1267626 0.0076041315 0.007329090 0.009971693 0.01339406
## 5
## 6
           6 0.8509641 0.0139793124 0.015195593 0.015170417 0.01434085
write.csv(out.jk.coefs,file=file.path(outdir,paste0(inVar,
                                              '_Bootstrap_PLSR_Coefficients.csv')),
        row.names=FALSE)
```

## Create core PLSR outputs

```
print(paste("Output directory: ", outdir))
## [1] "Output directory: /var/folders/xp/h3k9vf3n2jx181ts786_yjrn9c2gjq/T//RtmpKlCiie"
# 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,
                              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] "N_g_m2_Bootstrap_PLSR_Coefficients.csv"

## [2] "N_g_m2_Bootstrap_Regression_Coefficients.png"

## [3] "N_g_m2_Cal_PLSR_Dataset.csv"
```

```
[4] "N_g_m2_Cal_Val_Histograms.png"
##
   [5] "N_g_m2_Cal_Val_Scatterplots.png"
   [6] "N_g_m2_Cal_Val_Spectra.png"
##
##
   [7] "N_g_m2_Coefficient_VIP_plot.png"
    [8] "N_g_m2_Observed_PLSR_CV_Pred_15comp.csv"
##
   [9] "N_g_m2_PLSR_Coefficients_15comp.csv"
##
## [10] "N_g_m2_PLSR_Component_Selection.png"
## [11] "N_g_m2_PLSR_Validation_Scatterplot.png"
## [12] "N_g_m2_PLSR_VIPs_15comp.csv"
## [13] "N_g_m2_Val_PLSR_Dataset.csv"
## [14] "N_g_m2_Validation_PLSR_Pred_15comp.csv"
## [15] "N_g_m2_Validation_RMSEP_R2_by_Component.png"
```