

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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2022-03-14

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","spectratarait")
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/spectratarait

## 
## 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)

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

Load internal Ely et al 2019 dataset

```
data("ely_plsr_data")
head(ely_plsr_data) [,1:8]

##   Species_Code      Common_Name C_N_mass   C_g_m2 H20_g_m2 LMA_g_m2   N_g_m2
## 1      HEAN3 common sunflower    7.58 15.61210   167.63    36.40 2.103694
## 2      HEAN3 common sunflower    8.33 14.73724   164.68    34.65 1.231713
## 3      HEAN3 common sunflower    7.70 15.02495   156.95    35.08 1.764752
## 4      CUSA4  garden cucumber   7.40 11.14835   111.52    26.23 1.287963
## 5      CUSA4  garden cucumber   7.47 11.60735   123.58    26.71 1.411361
## 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/RtmpbpAmFt"
```

Full PLSR dataset

```
Start.wave <- 500
End.wave <- 2400
```

```
wv <- seq(Start.wave,End.wave,1)
plsr_data <- ely_plsr_data
head(plsr_data)[,1:6]

##   Species_Code      Common_Name C_N_mass   C_g_m2 H20_g_m2 LMA_g_m2
## 1       HEAN3 common sunflower    7.58 15.61210   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       CUSA4  garden cucumber    7.47 11.60735   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%
## OCBA    Cal: 68.421%
## POPUL   Cal: 71.429%
## GLMA4   Cal: 70.588%
## PHVU    Cal: 66.667%
names(split_data)
```

```
## [1] "cal_data" "val_data"
cal.plsr.data <- split_data$cal_data
head(cal.plsr.data)[1:8]
```

```
##   Species_Code      Common_Name C_N_mass   C_g_m2 H20_g_m2 LMA_g_m2   N_g_m2
## 1       HEAN3 common sunflower    7.58 15.61210   167.63    36.40 2.103694
## 2       HEAN3 common sunflower    8.33 14.73724   164.68    34.65 1.231713
## 4       CUSA4  garden cucumber    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    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
##   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
head(val.plsr.data)[1:8]

##   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
## 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=""))

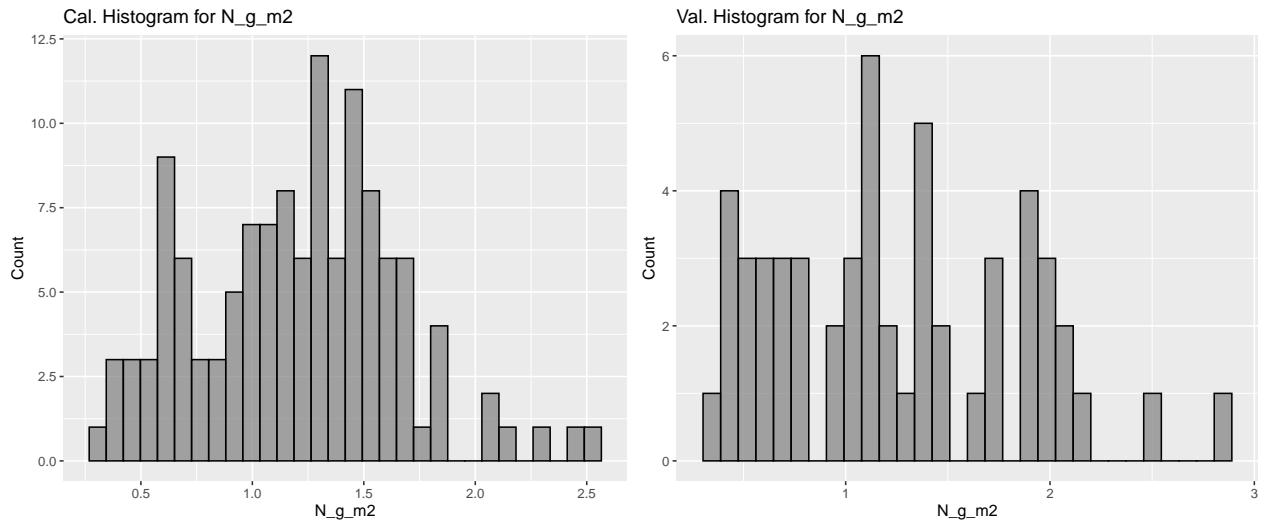
## [1] "Cal observations: 124"
print(paste("Val observations: ",dim(val.plsr.data)[1],sep=""))

## [1] "Val observations: 54"

cal_hist_plot <- qplot(cal.plsr.data[,paste0(inVar)],geom="histogram",
                        main = paste0("Cal. Histogram for ",inVar),
                        xlab = paste0(inVar),ylab = "Count",fill=I("grey50"),col=I("black"),
                        alpha=I(.7))
val_hist_plot <- qplot(val.plsr.data[,paste0(inVar)],geom="histogram",
                        main = paste0("Val. Histogram for ",inVar),
                        xlab = paste0(inVar),ylab = "Count",fill=I("grey50"),col=I("black"),
                        alpha=I(.7))
histograms <- grid.arrange(cal_hist_plot, val_hist_plot, ncol=2)

## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

```



```

ggsave(filename = file.path(outdir,paste0(inVar,"_Cal_Val_Histograms.png")),
       plot = histograms,
       device="png", width = 30,
       height = 12, units = "cm",
       dpi = 300)
# output cal/val data
write.csv(cal.plsr.data,file=file.path(outdir,paste0(inVar,'_Cal_PLSR_Dataset.csv')),
          row.names=FALSE)
write.csv(val.plsr.data,file=file.path(outdir,paste0(inVar,'_Val_PLSR_Dataset.csv')),
          row.names=FALSE)

```

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))])
cal.plsr.data <- data.frame(cal.plsr.data[, which(names(cal.plsr.data) %notin% paste0("Wave_",wv))],
                               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))])
val.plsr.data <- data.frame(val.plsr.data[, which(names(val.plsr.data) %notin% paste0("Wave_",wv))],
                               Spectra=I(val_spec))
head(val.plsr.data)[1:5]

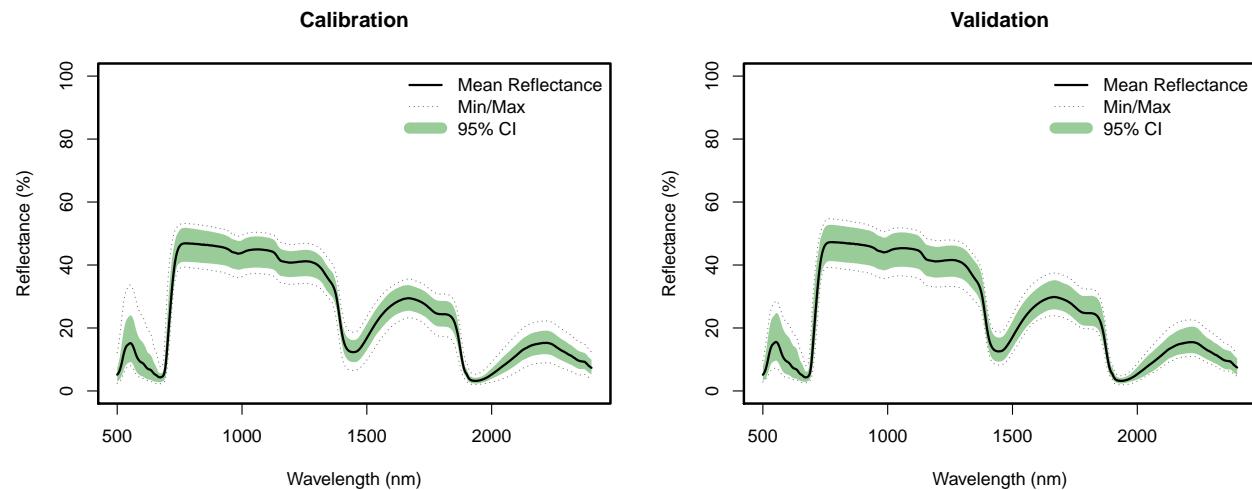
##   Species_Code      Common_Name C_N_mass   C_g_m2 H20_g_m2
## 3        HEAN3    common sunflower    7.70 15.024947   156.95
## 5       CUSA4     garden cucumber    7.47 11.607347   123.58
## 8        CUPE     field pumpkin    7.67 12.466238   124.67
## 9        CUPE     field pumpkin    7.64 17.100448   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
spectrattrait::f.plot.spec(Z=cal.plsr.data$Spectra,wv=wv,plot_label="Calibration")
spectrattrait::f.plot.spec(Z=val.plsr.data$Spectra,wv=wv,plot_label="Validation")
```



```
dev.copy(png,file.path(outdir,paste0(inVar,'_Cal_Val_Spectra.png')),
       height=2500,width=4900, res=340)
```

```
## quartz_off_screen
##                      3
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)
}

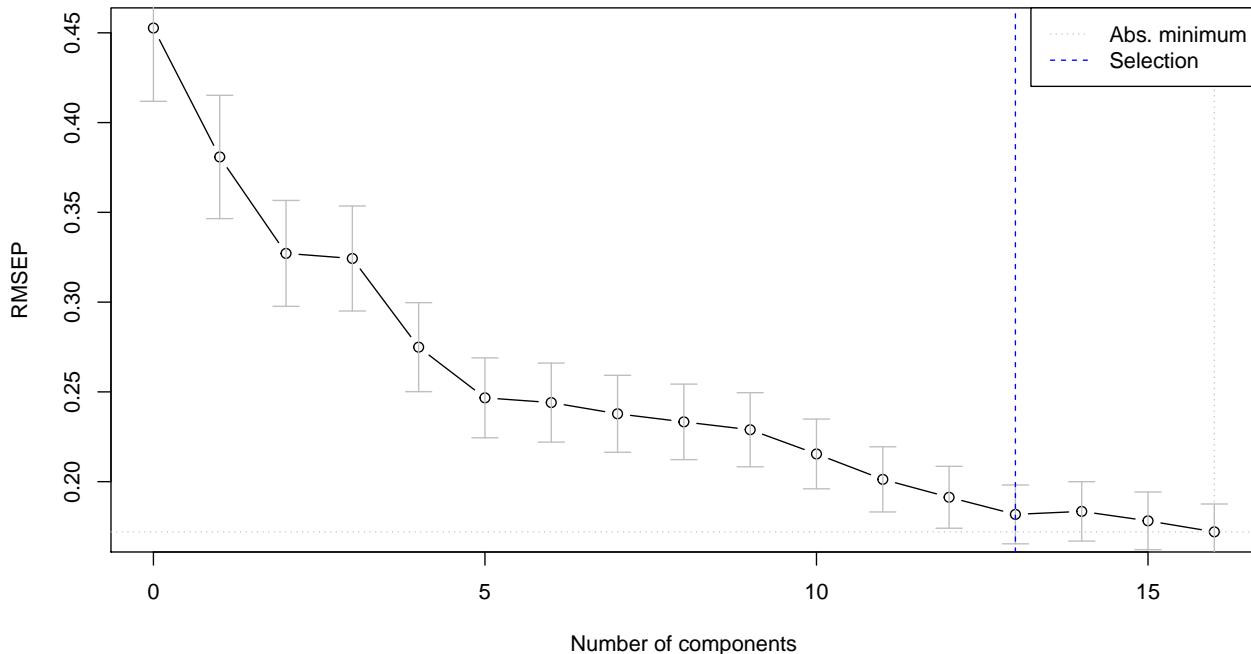
method <- "pls" #pls, firstPlateau, firstMin
random_seed <- 1245565
seg <- 50
maxComps <- 16
iterations <- 80
prop <- 0.70
```

```

if (method=="pls") {
  nComps <- spectratarait::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 <- spectratarait::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 ***"



```

## [1] "*** Optimal number of components: 13"
dev.copy(png,file.path(outdir,paste0(paste0(inVar, "_PLSR_Component_Selection.png"))),
        height=2800, width=3400, res=340)

```

```

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

## pdf
##      2

```

Fit final model

```

plsr.out <- plsr(as.formula(paste(inVar, "~", "Spectra")), scale=FALSE, ncomp=nComps, validation="LOO",
                  trace=FALSE, data=cal.plsr.data)
fit <- plsr.out$fitted.values[, 1:nComps]
pls.options(parallel = NULL)

# External validation fit stats
par(mfrow=c(1, 2)) # B, L, T, R
pls::RMSEP(plsr.out, newdata = val.plsr.data)

## (Intercept) 1 comps 2 comps 3 comps 4 comps 5 comps
## 0.5908     0.4735 0.4162 0.4037 0.3347 0.3023
## 6 comps     7 comps 8 comps 9 comps 10 comps 11 comps
## 0.2993     0.3081 0.2814 0.2445 0.2276 0.2104
## 12 comps    13 comps
## 0.1954     0.2003

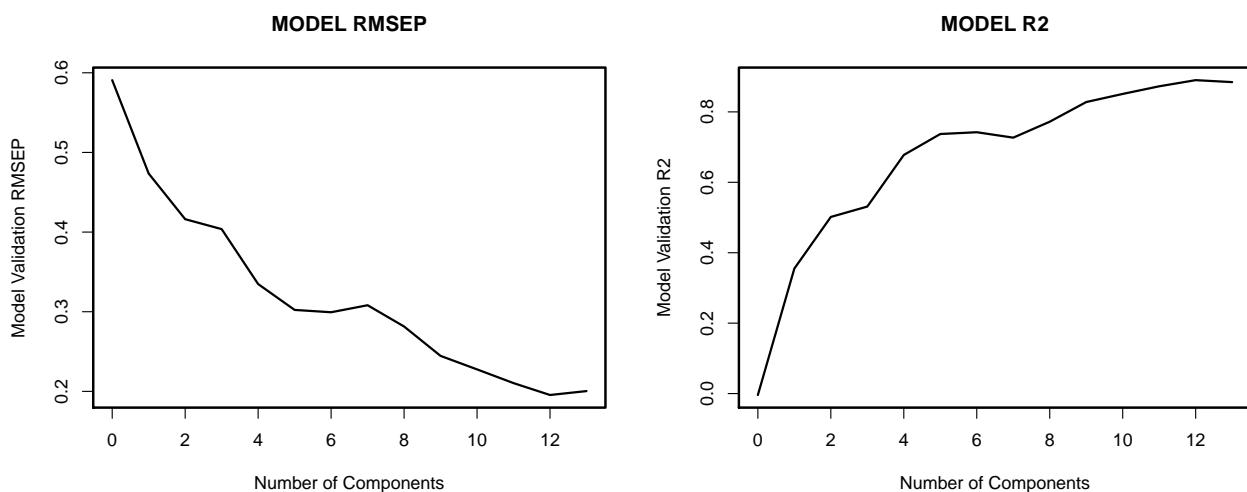
plot(pls::RMSEP(plsr.out, estimate=c("test"), newdata = val.plsr.data), main="MODEL RMSEP",
      xlab="Number of Components", ylab="Model Validation RMSEP", lty=1, col="black", cex=1.5, lwd=2)
box(lwd=2.2)

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
## 12 comps    13 comps
## 0.890124 0.884529

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(lwd=2.2)

```



```

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")],
                                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 Common_Name C_N_mass C_g_m2 H20_g_m2 LMA_g_m2 N_g_m2
## 1      HEAN3  common sunflower    7.58 15.61210   167.63   36.40 2.103694
## 2      HEAN3  common sunflower    8.33 14.73724   164.68   34.65 1.231713
## 4      CUSA4   garden cucumber    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     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      1.820666          1.702501       -0.40119317
## 2      1.609632          1.711772        0.48005882
## 4      1.364985          1.275526       -0.01243687
## 6      1.126062          1.060119       -0.05758587
## 7      1.227538          1.226708        0.01137583
## 10     1.358638          1.365181       0.06107105

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)

val.plsr.output <- data.frame(val.plsr.data[, which(names(val.plsr.data) %notin% "Spectra")],
                                 PLSR_Predicted=as.vector(predict(plsr.out,
                                                               newdata = val.plsr.data,
                                                               ncomp=nComps, type="response"))[, ,1]))
val.plsr.output <- val.plsr.output %>%
  mutate(PLSR_Residuals = PLSR_Predicted-get(inVar))
head(val.plsr.output)

##   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
## 15     OCBA     sweet basil    8.13 16.975969   173.30   38.65 1.1246459
##   PLSR_Predicted PLSR_Residuals
## 3      1.7125176   -0.052233917
## 5      1.4618447    0.050483171
## 8      1.0951891   -0.051652168

```

```

## 9      1.2152379   0.076220509
## 13     0.7992342   -0.149119020
## 15     1.1267054   0.002059572

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)

rng_quant <- quantile(cal.plsr.output[,inVar], probs = c(0.001, 0.999))
cal_scatter_plot <- ggplot(cal.plsr.output, aes(x=PLSR_CV_Predicted, y=get(inVar))) +
  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)) +
  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))

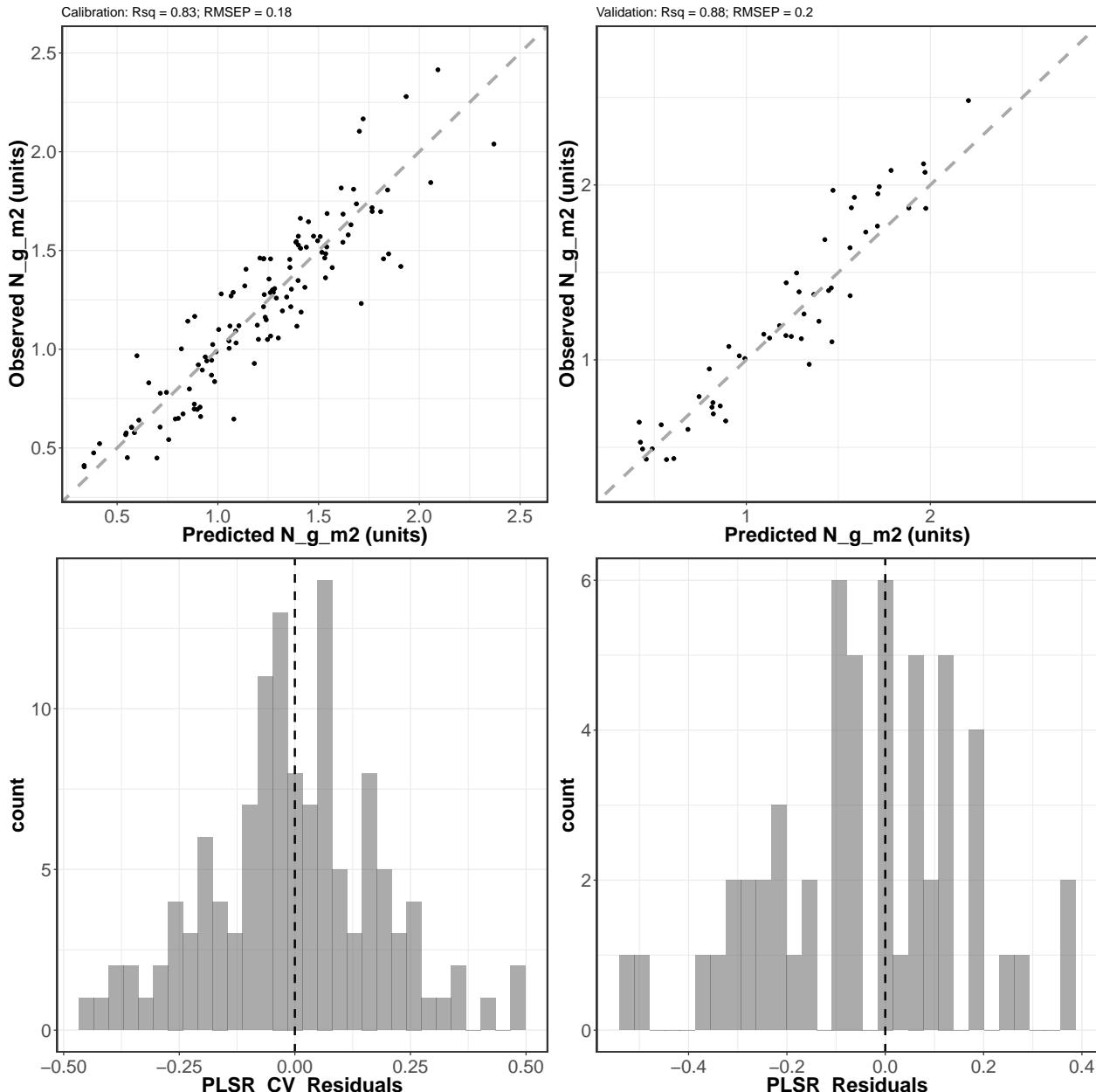
rng_quant <- quantile(val.plsr.output[,inVar], probs = c(0.001, 0.999))
val_scatter_plot <- ggplot(val.plsr.output, aes(x=PLSR_Predicted, y=get(inVar))) +
  theme_bw() + geom_point() + geom_abline(intercept = 0, slope = 1, color="dark grey",
                                           linetype="dashed", 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)) +
  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 3 rows containing missing values (geom_point).
Removed 3 rows containing missing values (geom_point).
`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.



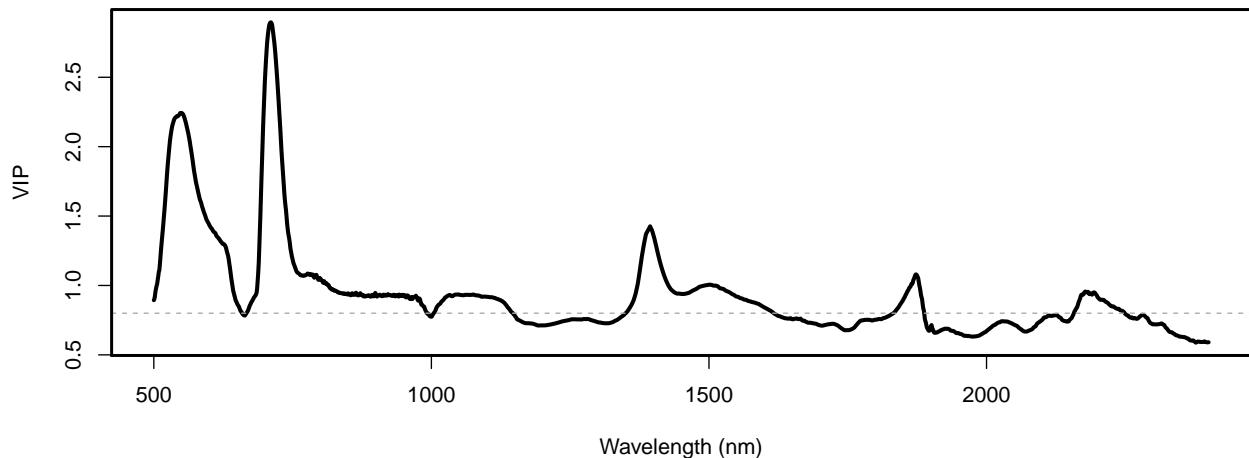
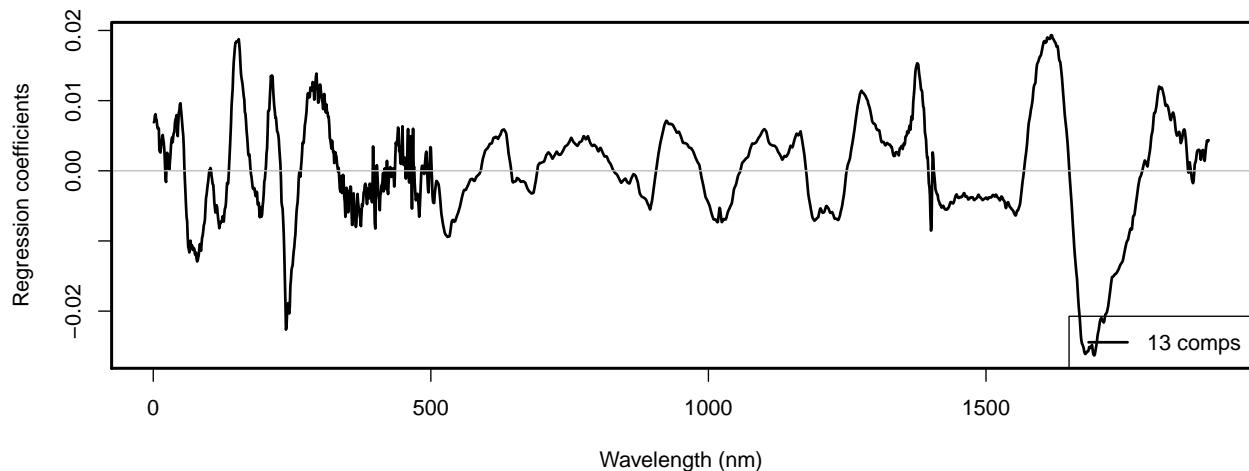
```
ggsave(filename = file.path(outdir, paste0(inVar, "_Cal_Val_Scatterplots.png")),
       plot = scatterplots, device="png",
       width = 32,
       height = 30, units = "cm",
```

```
dpi = 300)
```

Generate Coefficient and VIP plots

```
vips <- spectratarait::VIP(plsr.out)[nComps,]
par(mfrow=c(2,1))
plot(plsr.out, plottype = "coef", xlab="Wavelength (nm)",
     ylab="Regression coefficients", legendpos = "bottomright",
     ncomp=nComps, lwd=2)
box(lwd=2.2)
plot(seq(Start.wave,End.wave,1),vips,xlab="Wavelength (nm)",ylab="VIP",cex=0.01)
lines(seq(Start.wave,End.wave,1),vips,lwd=3)
abline(h=0.8,lty=2,col="dark grey")
box(lwd=2.2)
```

N_g_m2



```
dev.copy(png,file.path(outdir,paste0(inVar,'_Coefficient_VIP_plot.png')),
        height=3100, width=4100, res=340)
```

```

## 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
pls_r_permutation <- spectratarait::pls_permutation(dataset=cal.plsr.data, targetVariable=inVar,
                                                       maxComps=nComps,
                                                       iterations=iterations, prop=prop,
                                                       verbose = FALSE)

## [1] "*** Running permutation test. Please hang tight, this can take awhile ***"
## [1] "Options:"
## [1] "Max Components: 13 Iterations: 500 Data Proportion (percent): 70"
## [1] "*** Providing PRESS and coefficient array output ***"
bootstrap_intercept <- pls_r_permutation$coef_array[1,,nComps]
bootstrap_coef <- pls_r_permutation$coef_array[2:length(pls_r_permutation$coef_array[,1,nComps]), ,nComps]
rm(pls_r_permutation)

# apply coefficients to left-out validation data
interval <- c(0.025,0.975)
Bootstrap_Pred <- val.plsr.data$Spectra %*% bootstrap_coef +
  matrix(rep(bootstrap_intercept, length(val.plsr.data[,inVar])), byrow=TRUE,
        ncol=length(bootstrap_intercept))
Interval_Conf <- apply(X = Bootstrap_Pred, MARGIN = 1, FUN = quantile,
                       probs=c(interval[1], interval[2]))
sd_mean <- apply(X = Bootstrap_Pred, MARGIN = 1, FUN = sd)
sd_res <- sd(val.plsr.output$PLSR_Residuals)
sd_tot <- sqrt(sd_mean^2+sd_res^2)
val.plsr.output$LCI <- Interval_Conf[1,]
val.plsr.output$UCI <- Interval_Conf[2,]
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)

##   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
## 15      OCBA    sweet basil   8.13 16.975969 173.30 38.65 1.1246459
## PLSR_Predicted PLSR_Residuals      LCI      UCI      LPI      UPI
## 3       1.7125176 -0.052233917 1.5070086 1.8760564 1.2810247 2.144011
## 5       1.4618447  0.050483171 1.2909822 1.5475356 1.0541359 1.869553
## 8       1.0951891 -0.051652168 0.9595488 1.2335912 0.6846083 1.505770
## 9       1.2152379  0.076220509 1.0746965 1.3367675 0.8068229 1.623653
## 13      0.7992342 -0.149119020 0.6820207 0.9451323 0.3899050 1.208563
## 15      1.1267054  0.002059572 1.0316572 1.2737521 0.7209233 1.532488

```

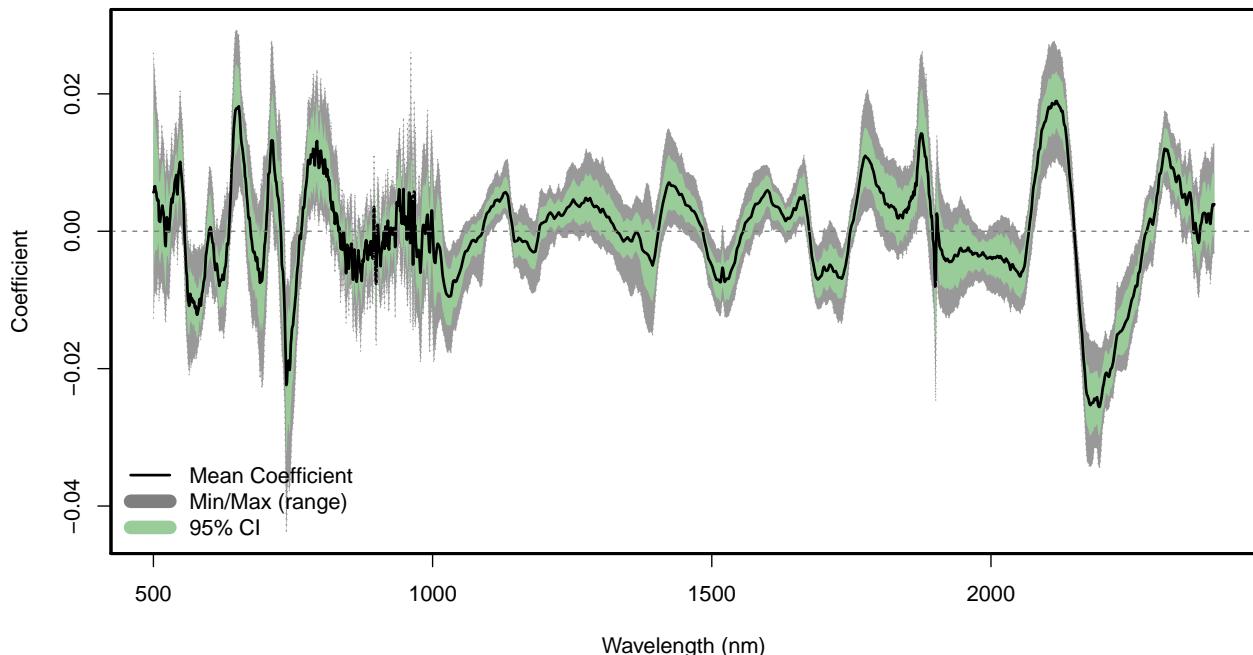
Jackknife coefficient plot

```

# Bootstrap regression coefficient plot
spectr trait::f.plot.coef(Z = t(bootstrap_coef), wv = wv,
                           plot_label="Bootstrap regression coefficients", position = 'bottomleft')
abline(h=0,lty=2,col="grey50")
box(lwd=2.2)

```

Bootstrap regression coefficients



```

dev.copy(png,file.path(outdir,paste0(inVar,'_Bootstrap_Regression_Coefficients.png')),
        height=2100, width=3800, res=340)

```

```

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

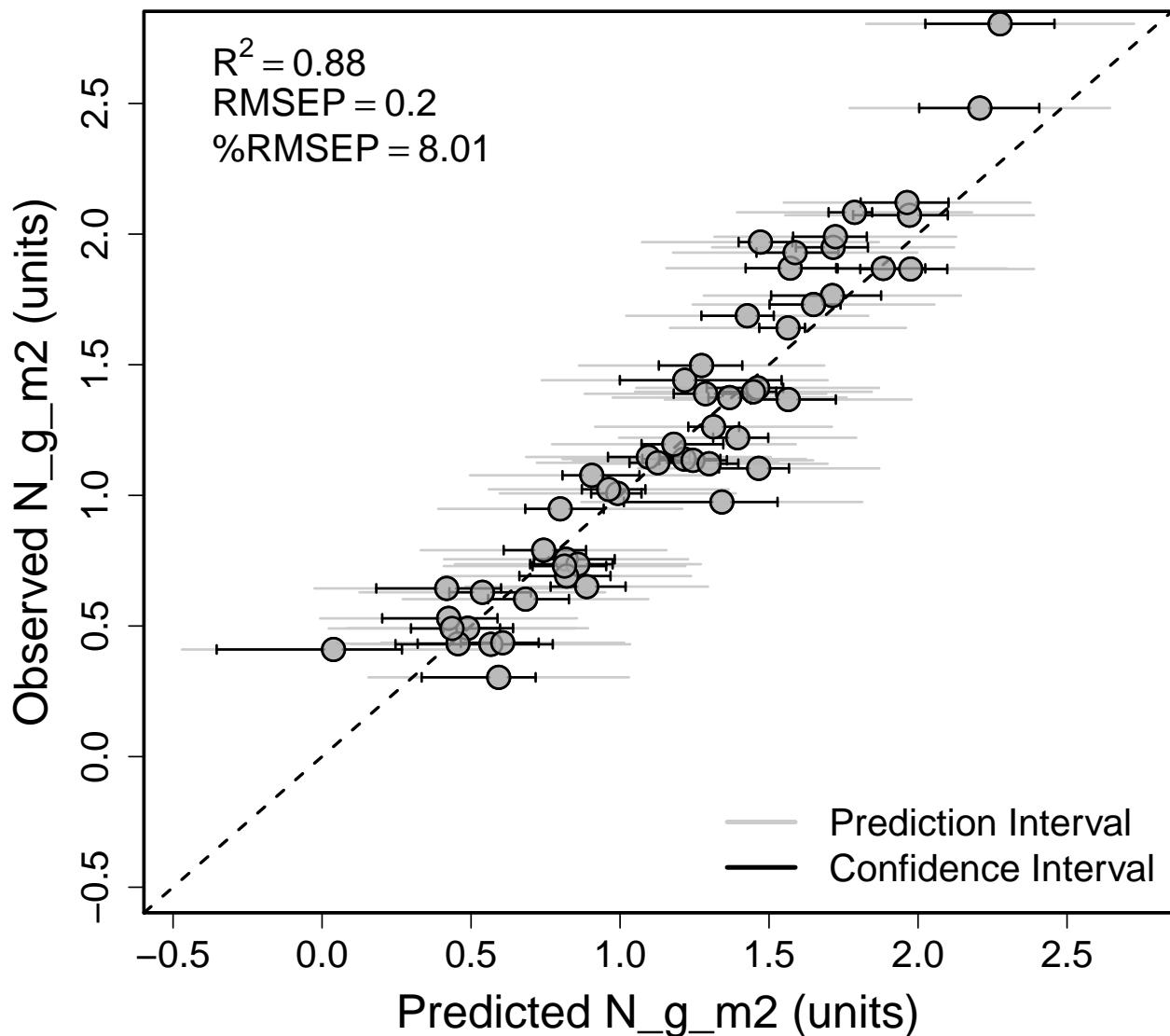
## pdf
##     2

```

Bootstrap validation plot

```
rmsep_percrmsep <- spectratrait::percent_rmse(plsr_dataset = val.plsr.output,
                                                inVar = inVar,
                                                residuals = val.plsr.output$PLSR_Residuals,
                                                range="full")

RMSEP <- rmsep_percrmsep$rmse
perc_RMSEP <- rmsep_percrmsep$perc_rmse
r2 <- round(pls::R2(plsr.out, newdata = val.plsr.data, intercept=F)$val[nComps],2)
expr <- vector("expression", 3)
expr[[1]] <- bquote(R^2==.(r2))
expr[[2]] <- bquote(RMSEP==.(round(RMSEP,2)))
expr[[3]] <- bquote("%RMSEP"==.(round(perc_RMSEP,2)))
rng_vals <- c(min(val.plsr.output$LPI), max(val.plsr.output$UPI))
par(mfrow=c(1,1), mar=c(4.2,5.3,1,0.4), oma=c(0, 0.1, 0, 0.2))
plotrix::plotCI(val.plsr.output$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(lwd=2.2)
```



```
dev.copy(png,file.path(outdir,paste0(inVar, "_PLSR_Validation_Scatterplot.png")),
       height=2800, width=3200, res=340)
```

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

## pdf
##      2
```

Output bootstrap results

```
# Bootstrap Coefficients
out.jk.coefs <- data.frame(Iteration=seq(1,length(bootstrap_intercept),1),
                           Intercept=bootstrap_intercept,t(bootstrap_coef))
names(out.jk.coefs) <- c("Iteration","Intercept",paste0("Wave_",wv))
head(out.jk.coefs)[1:6]
```

```

##   Iteration Intercept      Wave_500      Wave_501      Wave_502      Wave_503
## 1          1 -0.6617899 -0.0067918917 -0.006451152 -0.005571355 -0.004909648
## 2          2 -0.4636504 -0.0040384348 -0.001804902  0.001375426  0.002477500
## 3          3 -0.8146267  0.0031055624  0.003529288  0.005078394  0.005883173
## 4          4  0.7030872  0.0003883207  0.002887701  0.003961071  0.003223096
## 5          5  0.4765138  0.0045652557  0.005822813  0.005979498  0.004861387
## 6          6  0.4146289  0.0085296345  0.009692141  0.010451131  0.009300204

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//RtmpbpAmFt"
# 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)
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] "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_13comp.csv"
## [9] "N_g_m2_PLSR_Coefficients_13comp.csv"
## [10] "N_g_m2_PLSR_Component_Selection.png"
## [11] "N_g_m2_PLSR_Validation_Scatterplot.png"
## [12] "N_g_m2_PLSR_VIPs_13comp.csv"
## [13] "N_g_m2_Val_PLSR_Dataset.csv"
## [14] "N_g_m2_Validation_PLSR_Pred_13comp.csv"
## [15] "N_g_m2_Validation_RMSEP_R2_by_Component.png"
```