Spectra-trait PLSR example using leaf-level spectra and leaf nitrogen content (Narea, g/m2) data from 36 species growing in Rosa rugosa invaded coastal grassland communities in Belgium

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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 leaf nitrogen content (Narea, g/m2)

# **Getting Started**

#### Installation

```
## Loading required package: usethis
## Attaching package: 'remotes'
## The following objects are masked from 'package:devtools':
##
##
       dev_package_deps, install_bioc, install_bitbucket, install_cran,
       install_deps, install_dev, install_git, install_github,
##
##
       install_gitlab, install_local, install_svn, install_url,
       install_version, update_packages
##
  The following object is masked from 'package:usethis':
##
##
##
       git_credentials
##
## Attaching package: 'pls'
## The following object is masked from 'package:stats':
##
##
       loadings
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
## here() starts at /Users/sserbin/Data/GitHub/PLSR_for_plant_trait_prediction
```

```
##
## Attaching package: 'gridExtra'
## The following object is masked from 'package:dplyr':
##
## combine
```

# Setup other functions and options

```
### Setup other functions and options
github_dir <- file.path(here::here(), "R_Scripts")</pre>
source from gh <- TRUE
if (source_from_gh) {
  # Source helper functions from GitHub
  print("*** GitHub hash of functions.R file:")
  devtools::source_url("https://raw.githubusercontent.com/TESTgroup-BNL/PLSR_for_plant_trait_prediction
} else {
  functions <- file.path(github_dir, "functions.R")</pre>
  source(functions)
## [1] "*** GitHub hash of functions.R file:"
## SHA-1 hash of file is 2dc61e8f5adc6cd0473eb8f77d3d6c5f2d6a63ae
# not in
`%notin%` <- Negate(`%in%`)</pre>
# 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 <- "Narea_g_m2"</pre>
# What is the source dataset from EcoSIS?
ecosis id <- "9db4c5a2-7eac-4e1e-8859-009233648e89"
# 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] "/private/var/folders/xp/h3k9vf3n2jx181ts786\_yjrn9c2gjq/T/RtmpwGk127"

#### Grab data from EcoSIS

```
print(paste0("Output directory: ",getwd())) # check wd
## [1] "Output directory: /Users/sserbin/Data/GitHub/PLSR_for_plant_trait_prediction/vignettes"
dat_raw <- get_ecosis_data(ecosis_id = ecosis_id)</pre>
## [1] "**** Downloading Ecosis data ****"
## Downloading data...
## Parsed with column specification:
## cols(
##
     .default = col_double(),
##
     `Latin Species` = col_character(),
##
     ids = col_character(),
##
     `plot code` = col_character(),
##
     `species code` = col_character()
## )
## See spec(...) for full column specifications.
## Download complete!
head(dat raw)
## # A tibble: 6 x 2,164
##
     `Cw/EWT (cm3/cm~ `Latin Species` `Leaf area (mm2~ `Leaf calcium c~
##
                <dbl> <chr>
                                                                    <dbl>
                                                  <dbl>
## 1
              0.00887 Arrhenatherum ~
                                                   696.
                                                                   0.0291
              0.00824 Bromus sterilis
## 2
                                                   447.
                                                                   0.0230
## 3
              0.0280 Jacobaea vulga~
                                                  2418.
                                                                   0.0950
## 4
              0.0106 Rubus caesius
                                                  5719.
                                                                   0.0700
## 5
              0.00851 Arrhenatherum ~
                                                                   0.0286
                                                   671.
## 6
              0.0153 Crepis capilla~
                                                  1401.
                                                                   0.0470
## # ... with 2,160 more variables: `Leaf magnesium content per leaf area
       (mg/mm2) \ <dbl>, \Leaf mass per area (g/cm2) \ <dbl>, \Leaf nitrogen content
## #
       per leaf area (mg/mm2)` <dbl>, `Leaf phosphorus content per leaf area
## #
       (mg/mm2) \ <dbl>, \Leaf potassium content per leaf area (mg/mm2) \ <dbl>,
## #
       'Plant height vegetative (cm)' <dbl>, ids <chr>, 'plot code' <chr>,
## #
       `species code` <chr>, `350` <dbl>, `351` <dbl>, `352` <dbl>, `353` <dbl>,
       `354` <dbl>, `355` <dbl>, `356` <dbl>, `357` <dbl>, `358` <dbl>,
## #
## #
       `359` <dbl>, `360` <dbl>, `361` <dbl>, `362` <dbl>, `363` <dbl>,
       `364` <dbl>, `365` <dbl>, `366` <dbl>, `367` <dbl>, `368` <dbl>,
## #
       `369` <dbl>, `370` <dbl>, `371` <dbl>, `372` <dbl>, `373` <dbl>,
       `374` <dbl>, `375` <dbl>, `376` <dbl>, `377` <dbl>, `378` <dbl>,
## #
       `379` <dbl>, `380` <dbl>, `381` <dbl>, `382` <dbl>, `383` <dbl>,
## #
## #
       `384` <dbl>, `385` <dbl>, `386` <dbl>, `387` <dbl>, `388` <dbl>,
## #
       `389` <dbl>, `390` <dbl>, `391` <dbl>, `392` <dbl>, `393` <dbl>,
       `394` <dbl>, `395` <dbl>, `396` <dbl>, `397` <dbl>, `398` <dbl>,
## #
## #
       `399` <dbl>, `400` <dbl>, `401` <dbl>, `402` <dbl>, `403` <dbl>,
## #
       '404' <dbl>, '405' <dbl>, '406' <dbl>, '407' <dbl>, '408' <dbl>,
       `409` <dbl>, `410` <dbl>, `411` <dbl>, `412` <dbl>, `413` <dbl>,
## #
       `414` <dbl>, `415` <dbl>, `416` <dbl>, `417` <dbl>, `418` <dbl>,
## #
## #
       `419` <dbl>, `420` <dbl>, `421` <dbl>, `422` <dbl>, `423` <dbl>,
       `424` <dbl>, `425` <dbl>, `426` <dbl>, `427` <dbl>, `428` <dbl>,
## #
```

```
`429` <dbl>, `430` <dbl>, `431` <dbl>, `432` <dbl>, `433` <dbl>,
       `434` <dbl>, `435` <dbl>, `436` <dbl>, `437` <dbl>, `438` <dbl>,
       `439` <dbl>, `440` <dbl>, ...
names(dat_raw)[1:40]
    [1] "Cw/EWT (cm3/cm2)"
##
    [2] "Latin Species"
##
    [3] "Leaf area (mm2)"
   [4] "Leaf calcium content per leaf area (mg/mm2)"
   [5] "Leaf magnesium content per leaf area (mg/mm2)"
##
   [6] "Leaf mass per area (g/cm2)"
##
##
  [7] "Leaf nitrogen content per leaf area (mg/mm2)"
  [8] "Leaf phosphorus content per leaf area (mg/mm2)"
   [9] "Leaf potassium content per leaf area (mg/mm2)"
##
## [10] "Plant height vegetative (cm)"
## [11] "ids"
## [12] "plot code"
## [13] "species code"
## [14] "350"
## [15] "351"
## [16] "352"
## [17] "353"
## [18] "354"
## [19] "355"
## [20] "356"
## [21] "357"
## [22] "358"
## [23] "359"
## [24] "360"
## [25] "361"
## [26] "362"
## [27] "363"
## [28] "364"
## [29] "365"
## [30] "366"
## [31] "367"
## [32] "368"
## [33]
       "369"
## [34] "370"
## [35] "371"
## [36]
       "372"
       "373"
## [37]
## [38] "374"
## [39] "375"
## [40] "376"
```

## Create full 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])</pre>
```

```
colnames(Spectra) <- c(paste0("Wave_",wv))</pre>
sample_info <- dat_raw[,names(dat_raw) %notin% seq(350,2500,1)]</pre>
head(sample_info)
## # A tibble: 6 x 13
     `Cw/EWT (cm3/cm~ `Latin Species` `Leaf area (mm2~ `Leaf calcium c~
##
                <dbl> <chr>
                                                  <dbl>
                                                                   <dbl>
              0.00887 Arrhenatherum ~
                                                   696.
                                                                  0.0291
## 1
## 2
              0.00824 Bromus sterilis
                                                  447.
                                                                  0.0230
              0.0280 Jacobaea vulga~
## 3
                                                  2418.
                                                                  0.0950
              0.0106 Rubus caesius
## 4
                                                 5719.
                                                                  0.0700
## 5
              0.00851 Arrhenatherum ~
                                                  671.
                                                                  0.0286
              0.0153 Crepis capilla~
## 6
                                                  1401.
                                                                  0.0470
## # ... with 9 more variables: `Leaf magnesium content per leaf area
       (mg/mm2) ` <dbl>, `Leaf mass per area (g/cm2) ` <dbl>, `Leaf nitrogen content
       per leaf area (mg/mm2)` <dbl>, `Leaf phosphorus content per leaf area
       (mg/mm2) <dbl>, `Leaf potassium content per leaf area (mg/mm2) <dbl>,
## #
       `Plant height vegetative (cm)` <dbl>, ids <chr>, `plot code` <chr>,
## #
       `species code` <chr>
sample_info2 <- sample_info %>%
  select(Plant_Species=`Latin Species`,Species_Code=`species code`,Plot=`plot code`,
         Narea_mg_mm2=`Leaf nitrogen content per leaf area (mg/mm2)`)
sample_info2 <- sample_info2 %>%
# mutate(Narea_g_m2=Narea_mg_mm2*(0.001/1e-6)) # based on orig units should be this but conversion wro
 mutate(Narea_g_m2=Narea_mg_mm2*100) # this assumes orig units were g/mm2 or mg/cm2
head(sample_info2)
## # A tibble: 6 x 5
                           Species_Code Plot Narea_mg_mm2 Narea_g_m2
    Plant Species
##
     <chr>
                           <chr>
                                                                 <dbl>
                                        <chr>
                                                      <dbl>
## 1 Arrhenatherum elatius Arrela
                                        DC1
                                                    0.0126
                                                                 1 26
## 2 Bromus sterilis
                                        DC1
                                                    0.00682
                                                                 0.682
                           Broste
## 3 Jacobaea vulgaris
                                        DC1
                           Jacvul
                                                    0.0102
                                                                 1.02
## 4 Rubus caesius
                           Rubcae
                                        DC1
                                                    0.0121
                                                                 1.21
## 5 Arrhenatherum elatius Arrela
                                        DC2
                                                    0.0117
                                                                 1.17
                                        DC2
                                                    0.00877
## 6 Crepis capillaris
                           Creves
                                                                 0.877
plsr_data <- data.frame(sample_info2,Spectra)</pre>
rm(sample_info, sample_info2, Spectra)
plsr_data <- plsr_data[complete.cases(plsr_data[,names(plsr_data) %in%
```

Example data cleaning.

#### Create cal/val datasets

```
### Create cal/val datasets
## Make a stratified random sampling in the strata USDA_Species_Code and Domain
method <- "dplyr" #base/dplyr</pre>
```

c(inVar,paste0("Wave\_",wv))]),]

```
# base R - a bit slow
# dplyr - much faster
split data <- create data split(approach=method, split seed=1245565, prop=0.8,
                                group variables="Species Code")
names(split_data)
## [1] "cal data" "val data"
cal.plsr.data <- split_data$cal_data
head(cal.plsr.data)[1:8]
          Plant_Species Species_Code Plot Narea_mg_mm2 Narea_g_m2 Wave_500
                                      ZC3
                                                         3.240495 0.130885
## 1 Ammophila arenaria
                              Ammare
                                            0.03240495
## 2 Ammophila arenaria
                              Ammare
                                      MC2
                                            0.02806279
                                                         2.806279 0.135785
                                      ZC1
## 3 Ammophila arenaria
                              Ammare
                                            0.02041612
                                                         2.041612 0.147665
## 4 Ammophila arenaria
                              Ammare
                                     MC1
                                            0.02426549
                                                         2.426549 0.142765
                                      WC3
## 5 Ammophila arenaria
                              Ammare
                                            0.02807281
                                                         2.807281 0.151750
                              Ammare WR3
                                            0.02286678
                                                         2.286678 0.150850
## 6 Ammophila arenaria
    Wave 501 Wave 502
## 1 0.13175 0.132750
## 2 0.13685 0.138150
## 3 0.14910 0.150330
## 4 0.14390 0.145200
## 5 0.15275 0.154150
## 6 0.15185 0.152815
val.plsr.data <- split_data$val_data</pre>
head(val.plsr.data)[1:8]
##
            Plant_Species Species_Code Plot Narea_mg_mm2 Narea_g_m2
                                                                      Wave 500
## 184 Jacobaea vulgaris
                                Jacvul WC2 0.008756996 0.8756996 0.06736887
## 185 Potentilla reptans
                                Potrep WC2 0.010313464 1.0313464 0.07125000
## 186
            Rubus caesius
                                Rubcae WC2 0.007968454 0.7968454 0.05993560
                                Urtdio WC2 0.012737560 1.2737560 0.06508300
## 187
           Urtica dioica
## 188 Ammophila arenaria
                                Ammare WC3
                                             0.028072806 2.8072806 0.15175000
## 189
       Jacobaea vulgaris
                                Jacvul WC3 0.010251687 1.0251687 0.06805547
         Wave_501
##
                    Wave 502
## 184 0.06870667 0.07014220
## 185 0.07235000 0.07368350
## 186 0.06162000 0.06352233
## 187 0.06625000 0.06758350
## 188 0.15275000 0.15415000
## 189 0.06938000 0.07093553
rm(split data)
# Datasets:
print(paste("Cal observations: ",dim(cal.plsr.data)[1],sep=""))
## [1] "Cal observations: 183"
print(paste("Val observations: ",dim(val.plsr.data)[1],sep=""))
## [1] "Val observations: 73"
cal_hist_plot <- qplot(cal.plsr.data[,paste0(inVar)],geom="histogram",</pre>
                       main = paste0("Cal. Histogram for ",inVar),
```

```
xlab = paste0(inVar),ylab = "Count",fill=I("grey50"),col=I("black"),
                        alpha=I(.7))
val_hist_plot <- qplot(val.plsr.data[,paste0(inVar)],geom="histogram",</pre>
                        main = paste0("Val. Histogram for ",inVar),
                        xlab = 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`.
   Cal. Histogram for Narea_g_m2
                                                  Val. Histogram for Narea_g_m2
 20
                     Narea_g_m2
                                                                     Narea_g_m2
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))])</pre>
cal.plsr.data <- data.frame(cal.plsr.data[, which(names(cal.plsr.data) %notin% paste0("Wave_",wv))],</pre>
                            Spectra=I(cal_spec))
head(cal.plsr.data)[1:5]
          Plant_Species Species_Code Plot Narea_mg_mm2 Narea_g_m2
##
## 1 Ammophila arenaria
                              Ammare ZC3
                                             0.03240495
                                                          3.240495
## 2 Ammophila arenaria
                              Ammare MC2
                                             0.02806279
                                                          2.806279
                              Ammare ZC1
                                             0.02041612
                                                          2.041612
## 3 Ammophila arenaria
## 4 Ammophila arenaria
                                                          2.426549
                              Ammare MC1
                                             0.02426549
                                      WC3
## 5 Ammophila arenaria
                              Ammare
                                             0.02807281
                                                          2.807281
## 6 Ammophila arenaria
                              Ammare WR3
                                            0.02286678
                                                          2.286678
```

```
val_spec <- as.matrix(val.plsr.data[, which(names(val.plsr.data) %in% paste0("Wave_",wv))])</pre>
val.plsr.data <- data.frame(val.plsr.data[, which(names(val.plsr.data) %notin% paste0("Wave_",wv))],</pre>
                              Spectra=I(val_spec))
head(val.plsr.data)[1:5]
##
             Plant_Species Species_Code Plot Narea_mg_mm2 Narea_g_m2
## 184
        Jacobaea vulgaris
                                   Jacvul
                                           WC2
                                                 0.008756996
                                                               0.8756996
## 185 Potentilla reptans
                                   Potrep
                                           WC2
                                                 0.010313464
                                                               1.0313464
## 186
             Rubus caesius
                                   Rubcae
                                           WC2
                                                 0.007968454
                                                               0.7968454
             Urtica dioica
## 187
                                   Urtdio
                                           WC2
                                                 0.012737560
                                                               1.2737560
## 188 Ammophila arenaria
                                   Ammare
                                           WC3
                                                 0.028072806
                                                               2.8072806
## 189
        Jacobaea vulgaris
                                   Jacvul WC3
                                                0.010251687 1.0251687
plot cal and val spectra
par(mfrow=c(1,2)) # B, L, T, R
f.plot.spec(Z=cal.plsr.data$Spectra,wv=seq(Start.wave,End.wave,1),plot_label="Calibration")
f.plot.spec(Z=val.plsr.data$Spectra,wv=seq(Start.wave,End.wave,1),plot_label="Validation")
                      Calibration
                                                                          Validation
   9
                                                       8
                                   Mean Reflectance
                                                                                      Mean Reflectance
                                   95% CI
                                                                                      95% CI
   80
                                                       80
Reflectance (%)
                                                   Reflectance (%)
   9
                                                       9
   9
                                                       4
   20
                                                      20
       500
                1000
                          1500
                                    2000
                                                          500
                                                                    1000
                                                                             1500
                                                                                       2000
                      Wavelength (nm)
                                                                         Wavelength (nm)
dev.copy(png,file.path(outdir,paste0(inVar,'_Cal_Val_Spectra.png')),
         height=2500, width=4900, res=340)
## quartz off screen
##
dev.off();
## pdf
##
par(mfrow=c(1,1))
```

#### Use Jackknife permutation to determine optimal number of components

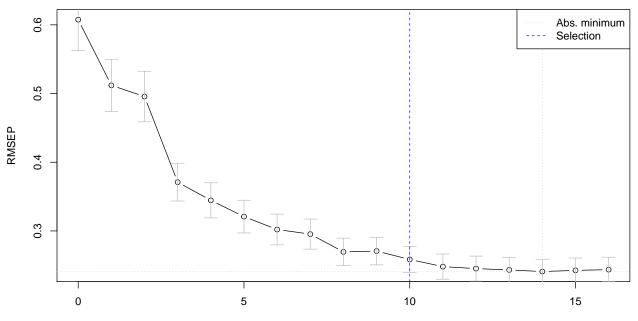
```
### Use permutation to determine the optimal number of components
if(grepl("Windows", sessionInfo()$running)){
  pls.options(parallel = NULL)
```

```
} else {
  pls.options(parallel = parallel::detectCores()-1)
method <- "pls" #pls, firstPlateau, firstMin</pre>
random_seed <- 1245565
seg <- 50
maxComps <- 16
iterations <- 80
prop <- 0.70
if (method=="pls") {
  # pls package approach - faster but estimates more components....
  nComps <- find_optimal_components(method=method, maxComps=maxComps, seg=seg,
                                     random_seed=random_seed)
  print(paste0("*** Optimal number of components: ", nComps))
} else {
  nComps <- find_optimal_components(dataset=cal.plsr.data, method=method, maxComps=maxComps,
                                     iterations=iterations, seg=seg, prop=prop,
                                    random_seed=random_seed)
}
```

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

##

2



Number of components

#### Fit final model

```
plsr.out <- plsr(as.formula(paste(inVar,"~","Spectra")),scale=FALSE,ncomp=nComps,validation="L00",
                   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
RMSEP(plsr.out, newdata = val.plsr.data)
## (Intercept)
                      1 comps
                                    2 comps
                                                  3 comps
                                                                 4 comps
                                                                               5 comps
        0.6346
                       0.5045
                                     0.4645
                                                    0.3415
                                                                  0.3296
                                                                                0.3037
##
##
       6 comps
                      7 comps
                                    8 comps
                                                  9 comps
                                                                10 comps
        0.2703
                       0.2659
                                     0.2524
                                                   0.2450
                                                                  0.2452
##
plot(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)
R2(plsr.out, newdata = val.plsr.data)
##
   (Intercept)
                      1 comps
                                    2 comps
                                                  3 comps
                                                                 4 comps
                                                                               5 comps
##
      -0.05977
                      0.33000
                                    0.43217
                                                  0.69298
                                                                 0.71415
                                                                               0.75732
##
       6 comps
                      7 comps
                                    8 comps
                                                  9 comps
                                                                10 comps
       0.80776
                      0.81389
                                    0.83228
                                                  0.84198
                                                                 0.84176
plot(R2(plsr.out,estimate=c("test"),newdata = val.plsr.data), main="MODEL R2",
     xlab="Number of Components", ylab="Model Validation R2", lty=1, col="black", cex=1.5, lwd=2)
box(1wd=2.2)
                     MODEL RMSEP
                                                                          MODEL R2
                                                       0.8
Model Validation RMSEP
                                                      9.0
                                                   Model Validation R2
   0.5
                                                       0.4
   0.4
   0.3
                                                       0.0
               2
        0
                              6
                                      8
                                             10
                                                                  2
                                                                                 6
                                                                                         8
                                                                                                10
                   Number of Components
                                                                       Number of Components
dev.copy(png,file.path(outdir,paste0(paste0(inVar,"_Validation_RMSEP_R2_by_Component.png"))),
          height=2800, width=4800, res=340)
## quartz_off_screen
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)
          Plant_Species Species_Code Plot Narea_mg_mm2 Narea_g_m2 PLSR_Predicted
##
## 1 Ammophila arenaria
                              Ammare ZC3
                                             0.03240495
                                                          3.240495
                                                                          2.672029
## 2 Ammophila arenaria
                                                                          2.651863
                              Ammare
                                      MC2
                                             0.02806279
                                                          2.806279
## 3 Ammophila arenaria
                              Ammare ZC1
                                             0.02041612
                                                          2.041612
                                                                          2.178056
## 4 Ammophila arenaria
                              Ammare
                                      MC1
                                             0.02426549
                                                          2.426549
                                                                          2.412013
## 5 Ammophila arenaria
                                      WC3
                                             0.02807281
                                                          2.807281
                                                                          2.452711
                              Ammare
## 6 Ammophila arenaria
                              Ammare WR3
                                             0.02286678
                                                          2.286678
                                                                          2.792340
##
     PLSR_CV_Predicted PLSR_CV_Residuals
## 1
              2.598245
                            -0.642250440
## 2
              2.652066
                            -0.154212969
## 3
              2.200588
                              0.158975634
                              0.009234491
## 4
              2.435784
## 5
              2.384049
                             -0.423231444
## 6
              2.943186
                             0.656508493
cal.R2 <- round(pls::R2(plsr.out)[[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")],
                              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 Species_Code Plot Narea_mg_mm2 Narea_g_m2 PLSR_Predicted
## 1
      Jacobaea vulgaris
                               Jacvul
                                      WC2
                                            0.008756996 0.8756996
                                                                         0.9462916
## 2 Potentilla reptans
                              Potrep
                                      WC2
                                            0.010313464
                                                         1.0313464
                                                                         1.5386676
## 3
          Rubus caesius
                              Rubcae
                                      WC2
                                            0.007968454
                                                         0.7968454
                                                                         0.8790482
## 4
          Urtica dioica
                              Urtdio
                                      WC2
                                            0.012737560
                                                        1.2737560
                                                                         1.1241560
## 5 Ammophila arenaria
                              Ammare
                                      WC3
                                            0.028072806 2.8072806
                                                                         2.4527108
      Jacobaea vulgaris
                              Jacvul WC3
                                            0.010251687 1.0251687
                                                                         1.1553688
     PLSR Residuals
##
## 1
         0.07059201
## 2
         0.50732119
## 3
         0.08220284
## 4
        -0.14959995
## 5
        -0.35456980
## 6
         0.13020008
```

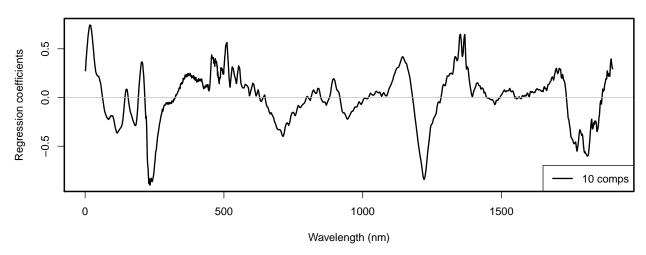
```
val.R2 <- round(pls::R2(plsr.out,newdata=val.plsr.data)[[1]][nComps],2)</pre>
val.RMSEP <- round(sqrt(mean(val.plsr.output$PLSR_Residuals^2)),2)</pre>
rng_quant <- quantile(cal.plsr.output[,inVar], probs = c(0.001, 0.999))</pre>
cal_scatter_plot <- ggplot(cal.plsr.output, aes(x=PLSR_CV_Predicted, y=get(inVar))) +</pre>
  theme_bw() + geom_point() + geom_abline(intercept = 0, slope = 1, color="dark grey",
                                           linetype="dashed", size=1.5) + xlim(rng_quant[1],
                                                                                rng quant[2]) +
  ylim(rng_quant[1], rng_quant[2]) +
  labs(x=paste0("Predicted ", paste(inVar), " (units)"),
       y=paste0("Observed ", paste(inVar), " (units)"),
       title=paste0("Calibration: ", paste0("Rsq = ", cal.R2), "; ", paste0("RMSEP = ",
                                                                              cal.RMSEP))) +
  theme(axis.text=element_text(size=18), legend.position="none",
        axis.title=element_text(size=20, face="bold"),
        axis.text.x = element_text(angle = 0, vjust = 0.5),
        panel.border = element_rect(linetype = "solid", fill = NA, size=1.5))
cal_resid_histogram <- ggplot(cal.plsr.output, aes(x=PLSR_CV_Residuals)) +</pre>
  geom_histogram(alpha=.5, position="identity") +
  geom_vline(xintercept = 0, color="black",
             linetype="dashed", size=1) + theme_bw() +
  theme(axis.text=element text(size=18), legend.position="none",
        axis.title=element_text(size=20, face="bold"),
        axis.text.x = element text(angle = 0, vjust = 0.5),
        panel.border = element_rect(linetype = "solid", fill = NA, size=1.5))
rng_quant <- quantile(val.plsr.output[,inVar], probs = c(0.001, 0.999))</pre>
val_scatter_plot <- ggplot(val.plsr.output, aes(x=PLSR_Predicted, y=get(inVar))) +</pre>
  theme_bw() + geom_point() + geom_abline(intercept = 0, slope = 1, color="dark grey",
                                           linetype="dashed", size=1.5) + xlim(rng_quant[1],
                                                                                rng_quant[2]) +
  ylim(rng_quant[1], rng_quant[2]) +
  labs(x=paste0("Predicted ", paste(inVar), " (units)"),
       y=paste0("Observed ", paste(inVar), " (units)"),
       title=paste0("Validation: ", paste0("Rsq = ", val.R2), "; ", paste0("RMSEP = ",
                                                                             val.RMSEP))) +
  theme(axis.text=element_text(size=18), legend.position="none",
        axis.title=element_text(size=20, face="bold"),
        axis.text.x = element_text(angle = 0, vjust = 0.5),
        panel.border = element_rect(linetype = "solid", fill = NA, size=1.5))
val_resid_histogram <- ggplot(val.plsr.output, aes(x=PLSR_Residuals)) +</pre>
  geom_histogram(alpha=.5, position="identity") +
  geom_vline(xintercept = 0, color="black",
             linetype="dashed", size=1) + theme_bw() +
  theme(axis.text=element_text(size=18), legend.position="none",
        axis.title=element_text(size=20, face="bold"),
        axis.text.x = element_text(angle = 0, vjust = 0.5),
        panel.border = element_rect(linetype = "solid", fill = NA, size=1.5))
# plot cal/val side-by-side
scatterplots <- grid.arrange(cal_scatter_plot, val_scatter_plot, cal_resid_histogram,
```

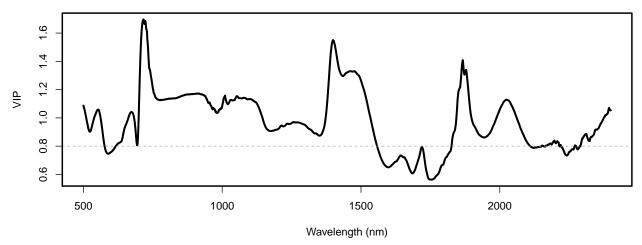
```
val_resid_histogram, nrow=2,ncol=2)
## Warning: Removed 2 rows containing missing values (geom_point).
## Warning: Removed 3 rows containing missing values (geom_point).
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
    Calibration: Rsq = 0.8; RMSEP = 0.26
                                                             Validation: Rsq = 0.84; RMSEP = 0.25
                                                          3.0
Observed Narea_g_m2 (units)
                                                       Observed Narea_g_m2 (units)
2.5
2.0.
                                                          1.0
              Predicted Narea_g_m2 (units)
                                                                       1.5 2.0 2.5
Predicted Narea_g_m2 (units)
                                                                                                       3.0
                                                          10.0
  30
                                                           7.5
  20
                                                       5.0
  10
                                                           2.5
   0
                                                           0.0
                   -0.5 0.0 PLSR_CV_Residuals
                                                                              PLSR_Residuals
          -1.0
                                             0.5
                                                                                                 0.3
                                                               -0.6
                                                                          -0.3
                                                                                                            0.6
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

# Narea\_g\_m2





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

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

#### Jackknife validation

```
if(grepl("Windows", sessionInfo()$running)){
  pls.options(parallel =NULL)
} else {
  pls.options(parallel = parallel::detectCores()-1)
jk.plsr.out <- pls::plsr(as.formula(paste(inVar, "~", "Spectra")), scale=FALSE,
                          center=TRUE, ncomp=nComps, validation="L00", trace=FALSE,
                          jackknife=TRUE,
                          data=cal.plsr.data)
pls.options(parallel = NULL)
Jackknife_coef <- f.coef.valid(plsr.out = jk.plsr.out, data_plsr = cal.plsr.data,</pre>
                                ncomp = nComps)
Jackknife_intercept <- Jackknife_coef[1,,,]</pre>
Jackknife_coef <- Jackknife_coef[2:dim(Jackknife_coef)[1],,,]</pre>
interval <-c(0.025, 0.975)
Jackknife_Pred <- val.plsr.data$Spectra %*% Jackknife_coef +</pre>
  matrix(rep(Jackknife_intercept, length(val.plsr.data[,inVar])), byrow=TRUE,
         ncol=length(Jackknife_intercept))
Interval_Conf <- apply(X = Jackknife_Pred, MARGIN = 1, FUN = quantile,</pre>
                        probs=c(interval[1], interval[2]))
sd_mean <- apply(X = Jackknife_Pred, MARGIN = 1, FUN =sd)</pre>
sd res <- sd(val.plsr.output$PLSR Residuals)</pre>
sd_tot <- sqrt(sd_mean^2+sd_res^2)</pre>
val.plsr.output$LCI <- Interval Conf[1,]</pre>
val.plsr.output$UCI <- Interval_Conf[2,]</pre>
val.plsr.output$LPI <- val.plsr.output$PLSR_Predicted-1.96*sd_tot
val.plsr.output$UPI <- val.plsr.output$PLSR_Predicted+1.96*sd_tot
head(val.plsr.output)
##
          Plant_Species_Code Plot Narea_mg_mm2 Narea_g_m2 PLSR_Predicted
```

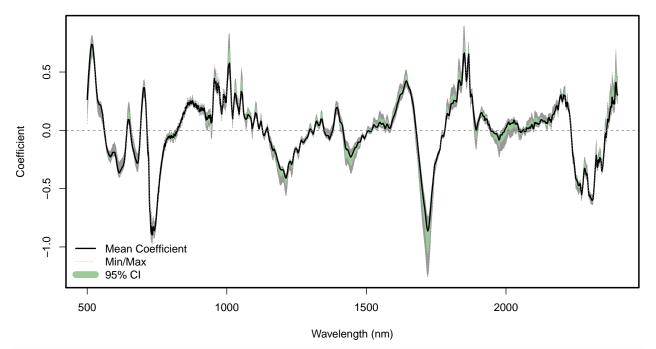
```
Jacvul WC2 0.008756996 0.8756996
## 1 Jacobaea vulgaris
                                                                      0.9462916
## 2 Potentilla reptans
                             Potrep WC2 0.010313464 1.0313464
                                                                      1.5386676
## 3
                             Rubcae WC2 0.007968454 0.7968454
         Rubus caesius
                                                                      0.8790482
## 4
         Urtica dioica
                             Urtdio WC2
                                          0.012737560 1.2737560
                                                                      1.1241560
## 5 Ammophila arenaria
                             Ammare WC3 0.028072806 2.8072806
                                                                      2.4527108
    Jacobaea vulgaris
                             Jacvul WC3 0.010251687 1.0251687
                                                                      1.1553688
                         LCI
                                   UCI
##
    PLSR_Residuals
                                             T.P.T
## 1
        0.07059201 0.9154961 0.9532972 0.4623162 1.430267
## 2
        0.50732119 1.4875834 1.5528063 1.0540777 2.023258
        0.08220284 0.8472007 0.9329303 0.3936085 1.364488
## 3
## 4
       -0.14959995 1.1075928 1.1743800 0.6395189 1.608793
## 5
       -0.35456980 2.4248448 2.5638131 1.9651833 2.940238
## 6
        0.13020008 1.1262731 1.1615642 0.6713762 1.639361
```

```
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)</pre>
```

```
##
          Plant_Species Species_Code Plot Narea_mg_mm2 Narea_g_m2 PLSR_Predicted
## 1
     Jacobaea vulgaris
                              Jacvul
                                      WC2
                                           0.008756996
                                                        0.8756996
                                                                        0.9462916
## 2 Potentilla reptans
                              Potrep
                                      WC2
                                           0.010313464 1.0313464
                                                                        1.5386676
## 3
          Rubus caesius
                              Rubcae
                                      WC2
                                           0.007968454 0.7968454
                                                                        0.8790482
## 4
          Urtica dioica
                              Urtdio
                                      WC2
                                           0.012737560
                                                        1.2737560
                                                                        1.1241560
## 5 Ammophila arenaria
                              Ammare
                                      WC3
                                           0.028072806
                                                         2.8072806
                                                                        2.4527108
## 6
     Jacobaea vulgaris
                              Jacvul
                                      WC3
                                           0.010251687
                                                        1.0251687
                                                                        1.1553688
##
     PLSR_Residuals
                          LCI
                                    UCI
                                               LPI
                                                        UPI
## 1
         0.07059201 0.9154961 0.9532972 0.4623162 1.430267
## 2
         0.50732119 1.4875834 1.5528063 1.0540777 2.023258
         0.08220284 0.8472007 0.9329303 0.3936085 1.364488
## 3
## 4
        -0.14959995 1.1075928 1.1743800 0.6395189 1.608793
## 5
        -0.35456980 2.4248448 2.5638131 1.9651833 2.940238
## 6
         0.13020008 1.1262731 1.1615642 0.6713762 1.639361
```

## Jackknife coefficient plot

### Jackknife regression coefficients



## quartz\_off\_screen

```
## 3
dev.off();
## pdf
## 2
```

#### Jackknife validation plot

```
RMSEP <- sqrt(mean(val.plsr.output$PLSR_Residuals^2))</pre>
pecr_RMSEP <- RMSEP/mean(val.plsr.output[,inVar])*100</pre>
r2 <- round(pls::R2(plsr.out, newdata = val.plsr.data)$val[nComps+1],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(pecr_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))
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]),
       err="x", pch=21, col="black", pt.bg=alpha("grey70",0.7), scol="grey50",
       cex=2, xlab=paste0("Predicted ", paste(inVar), " (units)"),
       ylab=paste0("Observed ", paste(inVar), " (units)"),
       cex.axis=1.5,cex.lab=1.8)
abline(0,1,1ty=2,1w=2)
legend("topleft", legend=expr, bty="n", cex=1.5)
box(1wd=2.2)
```

```
R^2 = 0.84
              RMSEP = 0.25
              %RMSEP = 15.42
     3.0
Observed Narea_g_m2 (units)
     2
                                      1.5
                                                2.0
                 0.5
                           1.0
                                                           2.5
                                                                     3.0
                                                                                3.5
                        Predicted Narea_g_m2 (units)
dev.copy(png,file.path(outdir,paste0(inVar,"_PLSR_Validation_Scatterplot.png")),
        height=2800, width=3200, res=340)
```

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

# Output jackknife results

```
out.jk.coefs <- data.frame(Iteration=seq(1,length(Jackknife_intercept),1),</pre>
                            Intercept=Jackknife_intercept,t(Jackknife_coef))
head(out.jk.coefs)[1:6]
```

```
Intercept Wave_500 Wave_501 Wave_502 Wave_503
        Iteration
                1 -0.001089661 0.3156927 0.3524556 0.3947195 0.4329382
## Seg 1
```

#### Create core PLSR outputs

```
print(paste("Output directory: ", outdir))
## [1] "Output directory: /var/folders/xp/h3k9vf3n2jx181ts786_yjrn9c2gjq/T//RtmpwGk127"
# 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: "

list.files(outdir)[grep(pattern = inVar, list.files(outdir))]

## [1] "Narea_g_m2_Cal_PLSR_Dataset.csv"

## [2] "Narea_g_m2_Cal_Val_Histograms.png"

## [3] "Narea_g_m2_Cal_Val_Scatterplots.png"

## [4] "Narea_g_m2_Cal_Val_Spectra.png"
```

```
## [5] "Narea_g_m2_Coefficient_VIP_plot.png"
## [6] "Narea_g_m2_Jackkife_PLSR_Coefficients.csv"
## [7] "Narea_g_m2_Jackknife_Regression_Coefficients.png"
## [8] "Narea_g_m2_Observed_PLSR_CV_Pred_10comp.csv"
## [9] "Narea_g_m2_PLSR_Coefficients_10comp.csv"
## [10] "Narea_g_m2_PLSR_Component_Selection.png"
## [11] "Narea_g_m2_PLSR_Validation_Scatterplot.png"
## [12] "Narea_g_m2_PLSR_VIPs_10comp.csv"
## [13] "Narea_g_m2_Val_PLSR_Dataset.csv"
## [14] "Narea_g_m2_Validation_PLSR_Pred_10comp.csv"
## [15] "Narea_g_m2_Validation_RMSEP_R2_by_Component.png"
```