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

Load libraries

```
list.of.packages <- c("pls", "dplyr", "reshape2", "here", "plotrix", "ggplot2", "gridExtra",</pre>
                       "spectratrait")
invisible(lapply(list.of.packages, library, character.only = TRUE))
##
## Attaching package: 'pls'
## The following object is masked from 'package:stats':
##
##
       loadings
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
## here() starts at /Users/sserbin/Data/GitHub/spectratrait
##
## Attaching package: 'gridExtra'
  The following object is masked from 'package:dplyr':
##
##
       combine
```

Setup other functions and options

```
### Setup other functions and options
# 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.q. "~/scratch/PLSR"
output_dir <- "tempdir"</pre>
```

Set working directory (scratch space)

[1] "/private/var/folders/xp/h3k9vf3n2jx181ts786_yjrn9c2gjq/T/Rtmp3BuJq0"

Grab data from EcoSIS

```
print(paste0("Output directory: ",getwd())) # check wd
## [1] "Output directory: /Users/sserbin/Data/GitHub/spectratrait/vignettes"
dat_raw <- spectratrait::get_ecosis_data(ecosis_id = ecosis_id)</pre>
## [1] "**** Downloading Ecosis data ****"
## Downloading data...
##
## cols(
    .default = col_double(),
##
    `Latin Species` = col_character(),
##
   ids = col_character(),
    `plot code` = col_character(),
    `species code` = col_character()
##
## )
## i Use `spec()` for the full column specifications.
```

Download complete!

```
head(dat_raw)
```

[14] "350" ## [15] "351" ## [16] "352"

```
## # A tibble: 6 x 2,164
     `Cw/EWT (cm3/cm2~ `Latin Species`
                                         `Leaf area (mm2~ `Leaf calcium content pe~
##
                 <dbl> <chr>
                                                     <dbl>
                                                                               <dbl>
## 1
               0.00887 Arrhenatherum el~
                                                      696.
                                                                              0.0291
## 2
               0.00824 Bromus sterilis
                                                      447.
                                                                              0.0230
               0.0280 Jacobaea vulgaris
                                                     2418.
                                                                              0.0950
               0.0106 Rubus caesius
## 4
                                                     5719.
                                                                              0.0700
## 5
               0.00851 Arrhenatherum el~
                                                      671.
                                                                              0.0286
## 6
               0.0153 Crepis capillaris
                                                     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)"
##
       "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"
```

```
## [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"
## [37] "373"
## [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)</pre>
Spectra <- as.matrix(dat_raw[,names(dat_raw) %in% wv])</pre>
colnames(Spectra) <- c(paste0("Wave_",wv))</pre>
sample_info <- dat_raw[,names(dat_raw) %notin% seq(350,2500,1)]</pre>
head(sample_info)
## # A tibble: 6 x 13
##
     `Cw/EWT (cm3/cm2~ `Latin Species`
                                          `Leaf area (mm2~ `Leaf calcium content pe~
##
                 <dbl> <chr>
                                                      <dbl>
                                                                                 <dbl>
                                                                                0.0291
## 1
               0.00887 Arrhenatherum el~
                                                       696.
               0.00824 Bromus sterilis
                                                       447.
                                                                                0.0230
               0.0280 Jacobaea vulgaris
## 3
                                                      2418.
                                                                                0.0950
## 4
               0.0106 Rubus caesius
                                                      5719.
                                                                                0.0700
## 5
               0.00851 Arrhenatherum el~
                                                                                0.0286
                                                       671.
## 6
               0.0153 Crepis capillaris
                                                      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
    Plant Species
                           Species_Code Plot Narea_mg_mm2 Narea_g_m2
                                        <chr>
##
     <chr>
                           <chr>
                                                     <dbl>
                                                                <dbl>
## 1 Arrhenatherum elatius Arrela
                                        DC1
                                                   0.0126
                                                                1.26
## 2 Bromus sterilis
                                                   0.00682
                                                                0.682
                           Broste
                                        DC1
## 3 Jacobaea vulgaris
                           Jacvul
                                        DC1
                                                   0.0102
                                                                1.02
## 4 Rubus caesius
                                        DC1
                           Rubcae
                                                   0.0121
                                                                1.21
## 5 Arrhenatherum elatius Arrela
                                        DC2
                                                   0.0117
                                                                1.17
## 6 Crepis capillaris
                           Creves
                                        DC2
                                                   0.00877
                                                                0.877
plsr_data <- data.frame(sample_info2,Spectra)</pre>
rm(sample_info,sample_info2,Spectra)
```

Example data cleaning.

Create cal/val datasets

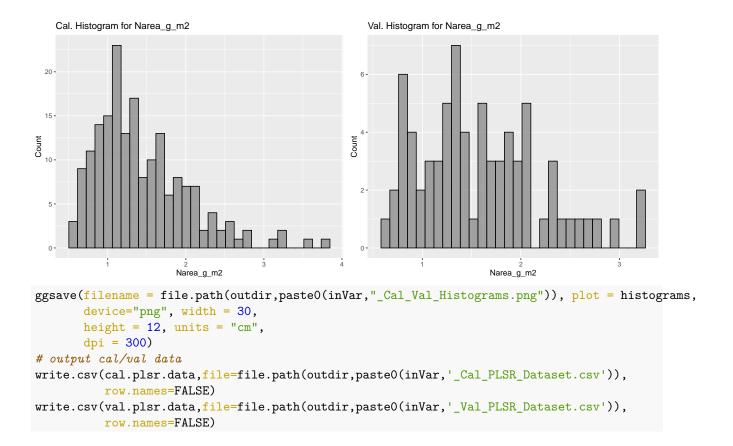
5 Ammophila arenaria

```
### Create cal/val datasets
## Make a stratified random sampling in the strata USDA_Species_Code and Domain
method <- "dplyr" #base/dplyr
# base R - a bit slow
# dplyr - much faster
split_data <- spectratrait::create_data_split(dataset=plsr_data, 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</pre>
head(cal.plsr.data)[1:8]
         Plant_Species Species_Code Plot Narea_mg_mm2 Narea_g_m2 Wave_500
## 1 Ammophila arenaria
                             Ammare ZC3
                                          0.03240495
                                                        3.240495 0.130885
## 2 Ammophila arenaria
                             Ammare MC2
                                          0.02806279 2.806279 0.135785
## 3 Ammophila arenaria
                             Ammare ZC1 0.02041612 2.041612 0.147665
                             Ammare MC1 0.02426549 2.426549 0.142765
## 4 Ammophila arenaria
```

0.02807281 2.807281 0.151750

Ammare WC3

```
## 6 Ammophila arenaria
                              Ammare WR3
                                            0.02286678
                                                         2.286678 0.150850
    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
                                Rubcae WC2 0.007968454 0.7968454 0.05993560
## 186
           Rubus caesius
## 187
            Urtica dioica
                                Urtdio WC2 0.012737560 1.2737560 0.06508300
## 188 Ammophila arenaria
                                Ammare WC3 0.028072806
                                                          2.8072806 0.15175000
       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[,pasteO(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 = pasteO("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`.
```



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
## 3 Ammophila arenaria
                              Ammare ZC1
                                             0.02041612
                                                          2.041612
## 4 Ammophila arenaria
                              Ammare
                                       MC1
                                             0.02426549
                                                          2.426549
## 5 Ammophila arenaria
                              Ammare WC3
                                             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
                                                            1.0313464
                                 Potrep
                                         WC2
                                              0.010313464
## 186
            Rubus caesius
                                 Rubcae
                                         WC2
                                              0.007968454
                                                           0.7968454
## 187
            Urtica dioica
                                 Urtdio
                                         WC2
                                              0.012737560
                                                           1.2737560
## 188 Ammophila arenaria
                                 Ammare WC3 0.028072806 2.8072806
```

plot cal and val spectra

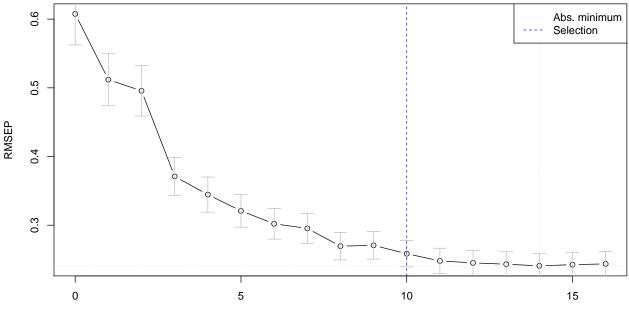
```
par(mfrow=c(1,2)) \# B, L, T, R
spectratrait::f.plot.spec(Z=cal.plsr.data$Spectra,wv=wv,plot_label="Calibration")
spectratrait::f.plot.spec(Z=val.plsr.data$Spectra,wv=wv,plot_label="Validation")
                         Calibration
                                                                                  Validation
    100
                                                             100
                                       Mean Reflectance
                                                                                               Mean Reflectance
                                                                                               Min/Max
                                       Min/Max
                                       95% CI
                                                                                               95% CI
    8
                                                            80
Reflectance (%)
                                                         Reflectance (%)
    9
                                                             9
    6
                                                            40
    20
                                                            20
                  1000
                             1500
                                                                           1000
                                                                                                 2000
        500
                                        2000
                                                                500
                                                                                      1500
                        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 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") {</pre>
```

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



Number of components

Fit final model

```
# 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
                                      0.4645
                                                    0.3415
                                                                                 0.3037
##
        0.6346
                       0.5045
                                                                   0.3296
       6 comps
                      7 comps
                                     8 comps
                                                   9 comps
                                                                 10 comps
##
        0.2703
                       0.2659
                                      0.2524
                                                    0.2450
                                                                   0.2452
##
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
                                                                                5 comps
                                                   3 comps
                                                                 4 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(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
                                                       0.8
    9.0
Model Validation RMSEP
                                                       9.0
                                                    Model Validation R2
    0.5
                                                       0.4
    0.4
                                                       0.2
    0.3
                                                       0.0
               2
                              6
                                      8
                                                                   2
                                                                                          8
                       4
                                             10
                                                                                  6
                                                                                                 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
                              Ammare MC2
                                            0.02806279
                                                          2.806279
                                                                         2.651863
                              Ammare ZC1
## 3 Ammophila arenaria
                                            0.02041612
                                                          2.041612
                                                                         2.178056
## 4 Ammophila arenaria
                              Ammare MC1
                                            0.02426549
                                                          2.426549
                                                                         2.412013
## 5 Ammophila arenaria
                              Ammare WC3
                                            0.02807281
                                                          2.807281
                                                                         2.452711
## 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
## 4
              2.435784
                             0.009234491
## 5
              2.384049
                            -0.423231444
              2.943186
                             0.656508493
cal.R2 <- round(pls::R2(plsr.out,intercept=F)[[1]][nComps],2)
cal.RMSEP <- round(sqrt(mean(cal.plsr.output$PLSR_CV_Residuals^2)),2)</pre>
val.plsr.output <- data.frame(val.plsr.data[, which(names(val.plsr.data) %notin% "Spectra")],</pre>
                              PLSR_Predicted=as.vector(predict(plsr.out,
                                                                newdata = val.plsr.data,
                                                                ncomp=nComps, type="response")[,,1]))
val.plsr.output <- val.plsr.output %>%
  mutate(PLSR_Residuals = PLSR_Predicted-get(inVar))
head(val.plsr.output)
            Plant_Species Species_Code Plot Narea_mg_mm2 Narea_g_m2 PLSR_Predicted
##
## 184 Jacobaea vulgaris
                                Jacvul WC2 0.008756996 0.8756996
                                                                          0.9462916
## 185 Potentilla reptans
                                        WC2 0.010313464
                                Potrep
                                                          1.0313464
                                                                          1.5386676
## 186
            Rubus caesius
                                Rubcae WC2 0.007968454
                                                           0.7968454
                                                                          0.8790482
## 187
            Urtica dioica
                                Urtdio WC2 0.012737560
                                                          1.2737560
                                                                          1.1241560
## 188 Ammophila arenaria
                                Ammare WC3 0.028072806
                                                          2.8072806
                                                                          2.4527108
       Jacobaea vulgaris
## 189
                                Jacvul WC3 0.010251687 1.0251687
                                                                          1.1553688
##
       PLSR_Residuals
## 184
           0.07059201
## 185
           0.50732119
## 186
           0.08220284
## 187
         -0.14959995
## 188
          -0.35456980
## 189
           0.13020008
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>
```

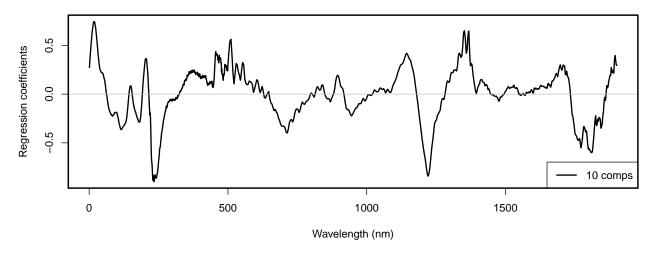
```
cal_scatter_plot <- ggplot(cal.plsr.output, aes(x=PLSR_CV_Predicted, y=get(inVar))) +</pre>
  theme_bw() + geom_point() + geom_abline(intercept = 0, slope = 1, color="dark grey",
                                           linetype="dashed", size=1.5) + xlim(rng_quant[1],
                                                                               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 2 rows containing missing values (geom_point).
## Warning: Removed 3 rows containing missing values (geom_point).
```

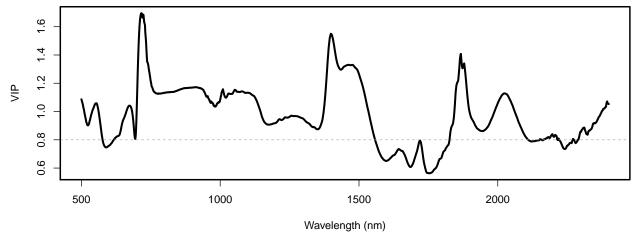
```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
      Calibration: Rsq = 0.82; RMSEP = 0.26
                                                                           Validation: Rsq = 0.84; RMSEP = 0.25
                                                                       3.0
                                                                    Observed Narea_g_m2 (units)
9.7
Observed Narea_g_m2 (units)
                  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
                        PLSR_CV_Residuals
                                                                                               PLSR_Residuals
             -1.0
                                                        0.5
                                                                             -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

```
vips <- spectratrait::VIP(plsr.out)[nComps,]
par(mfrow=c(2,1))</pre>
```

Narea_g_m2





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

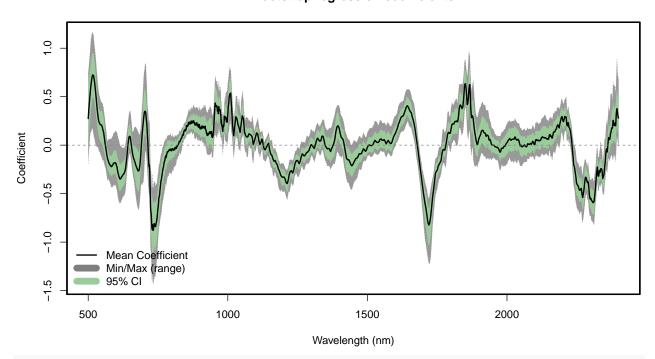
pdf ## 2

Bootstrap validation

```
if(grepl("Windows", sessionInfo()$running)){
  pls.options(parallel =NULL)
} else {
  pls.options(parallel = parallel::detectCores()-1)
### PLSR bootstrap permutation uncertainty analysis
iterations <- 500
                     # how many permutation iterations to run
                      # fraction of training data to keep for each iteration
prop <- 0.70
plsr_permutation <- spectratrait::pls_permutation(dataset=cal.plsr.data, maxComps=nComps,
                                                   iterations=iterations, prop=prop,
                                                   verbose = FALSE)
## [1] "*** Running permutation test. Please hang tight, this can take awhile ***"
## [1] "Options:"
## [1] "Max Components: 10 Iterations: 500 Data Proportion (percent): 70"
## [1] "*** Providing PRESS and coefficient array output ***"
bootstrap_intercept <- plsr_permutation$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$Spectra %*% bootstrap_coef +</pre>
  matrix(rep(bootstrap_intercept, length(val.plsr.data[,inVar])), byrow=TRUE,
         ncol=length(bootstrap intercept))
Interval_Conf <- apply(X = Bootstrap_Pred, MARGIN = 1, FUN = quantile,</pre>
                       probs=c(interval[1], interval[2]))
sd_mean <- apply(X = Bootstrap_Pred, MARGIN = 1, FUN = sd)</pre>
sd_res <- sd(val.plsr.output$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</pre>
head(val.plsr.output)
            Plant_Species Species_Code Plot Narea_mg_mm2 Narea_g_m2 PLSR_Predicted
## 184 Jacobaea vulgaris
                                 Jacvul WC2 0.008756996 0.8756996
                                                                           0.9462916
## 185 Potentilla reptans
                                Potrep WC2 0.010313464 1.0313464
                                                                           1.5386676
## 186
                                Rubcae WC2 0.007968454 0.7968454
            Rubus caesius
                                                                           0.8790482
## 187
            Urtica dioica
                                Urtdio WC2 0.012737560 1.2737560
                                                                           1.1241560
                                 Ammare WC3 0.028072806 2.8072806
## 188 Ammophila arenaria
                                                                           2.4527108
## 189
       Jacobaea vulgaris
                                 Jacvul WC3 0.010251687 1.0251687
                                                                           1.1553688
##
       PLSR Residuals
                            LCI
                                     UCI
                                                LPI
                                                         UPI
## 184
           0.07059201 0.8915898 1.008806 0.4588988 1.433684
           0.50732119\ 1.4007173\ 1.636452\ 1.0403747\ 2.036960
## 185
## 186
           0.08220284 0.6861219 1.160030 0.3405908 1.417506
## 187
          -0.14959995 0.9651982 1.245464 0.6198291 1.628483
          -0.35456980 2.1911406 2.619696 1.9245720 2.980850
## 188
```

Jackknife coefficient plot

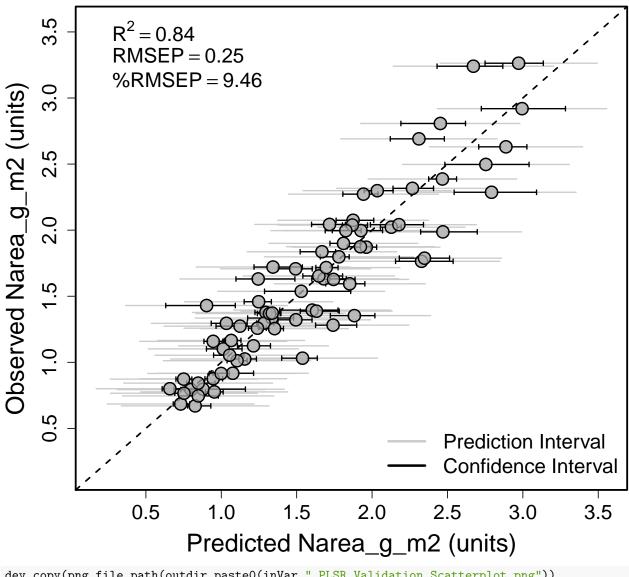
Bootstrap regression coefficients



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

Bootstrap validation plot

```
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)
```



Output bootstrap results

```
Iteration Intercept Wave_500 Wave_501 Wave_502 Wave_503
## 1
            1 -0.13686765 0.29141489 0.3287594 0.3654695 0.3999712
## 2
            2 -0.17612080 0.24114488 0.2827001 0.3182992 0.3552505
## 3
            3 0.34135463 0.21939317 0.2562451 0.2984578 0.3330333
## 4
            4 0.01511507 0.09404839 0.1299058 0.1792805 0.2240698
## 5
            5 0.06136605 0.12835311 0.1662322 0.2116938 0.2539193
## 6
            6 0.10925409 0.28154095 0.3107509 0.3539714 0.3932620
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//Rtmp3BuJqO"
# 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] "Narea_g_m2_Bootstrap_PLSR_Coefficients.csv"

## [2] "Narea_g_m2_Bootstrap_Regression_Coefficients.png"

## [3] "Narea_g_m2_Cal_PLSR_Dataset.csv"
```

[4] "Narea_g_m2_Cal_Val_Histograms.png" ## [5] "Narea_g_m2_Cal_Val_Scatterplots.png" [6] "Narea_g_m2_Cal_Val_Spectra.png" ## ## [7] "Narea_g_m2_Coefficient_VIP_plot.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"