Untitled3

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R. Markdown

This is an R Markdown document. Markdown is a simple formatting syntax for authoring HTML, PDF, and MS Word documents. For more details on using R Markdown see http://rmarkdown.rstudio.com.

When you click the **Knit** button a document will be generated that includes both content as well as the output of any embedded R code chunks within the document. You can embed an R code chunk like this:

```
Sys.time()
```

```
## [1] "2022-05-05 10:57:13 PDT"
library(tinytex)
library(tidyverse)
library(cowplot)
library(Cairo)
library(modelr)
library(gridExtra)
library(mixtools)
library(nlme)
library(car)
library(emmeans)
library(MuMIn)
library(ggpmisc)
library(gtable)
library(grid)
library(RColorBrewer)
library(segmented)
library(data.table)
library(scales)
library(sm)
library(rcompanion)
library(nlstools)
```

DATA

GreenSeeker NDVI Data

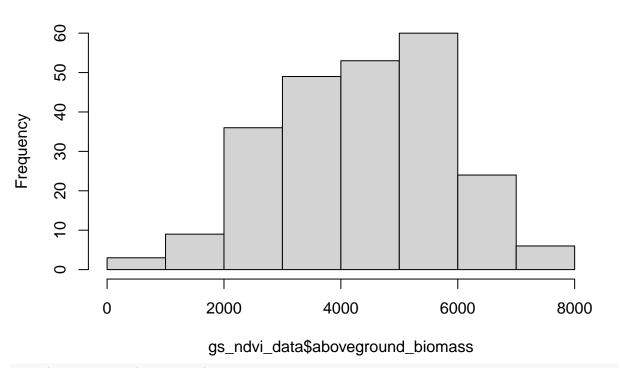
```
##
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
str(gs_ndvi_data , give.attr = FALSE)
## spec_tbl_df [328 x 17] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
## $ site_year
                      : chr [1:328] "Arbuckle-15" "Arbuckle-15" "Arbuckle-15" "Arbuckle-15" ...
                      : num [1:328] 2015 2015 2015 2015 ...
## $ year
## $ exp_plot_number : num [1:328] 101 102 103 104 105 201 202 203 204 205 ...
## $ block
                      : num [1:328] 1 1 1 1 1 2 2 2 2 2 ...
                      : num [1:328] 1 2 3 4 5 1 2 3 4 5 ...
## $ plot
## $ N level
                      : num [1:328] 125 225 0 75 175 0 175 125 225 75 ...
                    : num [1:328] 125 225 0 75 175 0 175 125 225 75 ...
## $ N_level_kgha
## $ biomass_plus_bag_g: num [1:328] 414 472 281 386 455 304 402 322 418 336 ...
## $ paper_bag_g
                   : num [1:328] 45 45 45 45 45 45 45 45 45 ...
## $ ring_size
                      ## $ num_of_paper_bags : num [1:328] 1 1 1 1 1 1 1 1 1 1 ...
## $ sample_weight_mg : num [1:328] 4.84 5.12 4.78 5.15 4.93 ...
## $ sample_N_ug
                      : num [1:328] 117.1 153.4 64.9 92.9 116 ...
                      : chr [1:328] "0.77" "0.82" "0.56" "0.72" ...
## $ NDVI_1
## $ NDVI 2
                      : chr [1:328] "n/a" "n/a" "n/a" "n/a" ...
## $ NDVI 3
                      : chr [1:328] "n/a" "n/a" "n/a" "n/a" ...
## $ NDVI 4
                      : chr [1:328] "n/a" "n/a" "n/a" "n/a" ...
gs_ndvi_data <- gs_ndvi_data %>%
 filter(!year %in% c("2015", "2016") & N_level_kgha != 275) #remove the years we don't need for this
str(gs_ndvi_data , give.attr = FALSE)
## spec_tbl_df [240 x 17] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
                      : chr [1:240] "Nicolaus-17" "Nicolaus-17" "Nicolaus-17" "Nicolaus-17" ...
## $ site_year
                      : num [1:240] 2017 2017 2017 2017 ...
## $ year
## $ exp_plot_number : num [1:240] 101 102 103 104 106 107 202 203 204 205 ...
## $ block
                      : num [1:240] 1 1 1 1 1 1 2 2 2 2 ...
## $ plot
                      : num [1:240] 1 2 3 4 6 7 2 3 4 5 ...
                      : num [1:240] 225 0 125 175 45 75 75 125 0 175 ...
## $ N_level
## $ N_level_kgha
                     : num [1:240] 225 0 125 175 45 75 75 125 0 175 ...
## $ biomass_plus_bag_g: num [1:240] 361 264 318 360 285 ...
## $ paper_bag_g
                     ## $ ring_size
                      : num [1:240] 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 ...
## $ num_of_paper_bags : num [1:240] 1 1 1 1 1 1 1 1 1 1 ...
## $ sample_weight_mg : num [1:240] 3.17 3.42 3.09 3.07 3.2 ...
## $ sample_N_ug
                      : num [1:240] 79.4 61.1 80.7 95.9 63.6 ...
## $ NDVI_1
                      : chr [1:240] "0.77" "0.50" "0.67" "0.79" ...
## $ NDVI 2
                      : chr [1:240] "0.77" "0.52" "0.67" "0.78" ...
## $ NDVI_3
                      : chr [1:240] "0.79" "0.46" "0.71" "0.79" ...
## $ NDVI 4
                      : chr [1:240] "n/a" "n/a" "n/a" "n/a" ...
gs_ndvi_data \leftarrow gs_ndvi_data[c(1:240), c(1:17)] #removes the empty rows and columns from the data frame
gs_ndvi_data$block <- factor(gs_ndvi_data$block)</pre>
gs_ndvi_data$year <- factor(gs_ndvi_data$year)</pre>
gs_ndvi_data$plot <- factor(gs_ndvi_data$plot)</pre>
gs_ndvi_data$N_level_kgha_f <- factor(gs_ndvi_data$N_level_kgha)</pre>
gs_ndvi_data$exp_plot_number <- factor(gs_ndvi_data$exp_plot_number)</pre>
```

```
gs_ndvi_data$site_year <- factor(gs_ndvi_data$site_year , levels = c( "Nicolaus-17" , "Williams-17" , ".
gs_ndvi_data$NDVI_1 <- as.numeric(as.character(gs_ndvi_data$NDVI_1))</pre>
## Warning: NAs introduced by coercion
gs_ndvi_data$NDVI_2 <- as.numeric(as.character(gs_ndvi_data$NDVI_2))</pre>
## Warning: NAs introduced by coercion
gs_ndvi_data$NDVI_3 <- as.numeric(as.character(gs_ndvi_data$NDVI_3))</pre>
gs_ndvi_data$NDVI_4 <- as.numeric(as.character(gs_ndvi_data$NDVI_4)) #qets the data right
## Warning: NAs introduced by coercion
str(gs_ndvi_data , give.attr = FALSE)
## tibble [240 x 18] (S3: tbl_df/tbl/data.frame)
## $ site_year
                      : Factor w/ 10 levels "Nicolaus-17",...: 1 1 1 1 1 1 1 1 1 1 ...
## $ year
                      : Factor w/ 3 levels "2017", "2018", ...: 1 1 1 1 1 1 1 1 1 1 ...
## $ exp_plot_number : Factor w/ 27 levels "101","102","103",..: 1 2 3 4 6 7 9 10 11 12 ...
                      : Factor w/ 4 levels "1","2","3","4": 1 1 1 1 1 1 2 2 2 2 ...
## $ block
## $ plot
                      : Factor w/ 7 levels "1","2","3","4",..: 1 2 3 4 6 7 2 3 4 5 ...
## $ N_level
                      : num [1:240] 225 0 125 175 45 75 75 125 0 175 ...
## $ N_level_kgha : num [1:240] 225 0 125 175 45 75 75 125 0 175 ...
## $ biomass_plus_bag_g: num [1:240] 361 264 318 360 285 ...
## $ paper_bag_g
                   ## $ ring_size
                      ## $ num_of_paper_bags : num [1:240] 1 1 1 1 1 1 1 1 1 1 ...
## $ sample_weight_mg : num [1:240] 3.17 3.42 3.09 3.07 3.2 ...
                      : num [1:240] 79.4 61.1 80.7 95.9 63.6 ...
## $ sample_N_ug
## $ NDVI 1
                      : num [1:240] 0.77 0.5 0.67 0.79 0.68 0.63 0.66 0.68 0.54 0.77 ...
## $ NDVI_2
                      : num [1:240] 0.77 0.52 0.67 0.78 0.69 0.56 0.65 0.68 0.5 0.75 ...
## $ NDVI 3
                      : num [1:240] 0.79 0.46 0.71 0.79 0.68 0.53 0.63 0.64 0.5 0.77 ...
## $ NDVI_4
                      : num [1:240] NA ...
                      : Factor w/ 11 levels "0", "45", "75", ...: 10 1 5 8 2 3 3 5 1 8 ...
## $ N_level_kgha_f
gs_ndvi_data <- gs_ndvi_data %>%
 mutate( biomass_dry_wt = biomass_plus_bag_g - (paper_bag_g * num_of_paper_bags) ,
         aboveground_biomass = (biomass_dry_wt / ring_size) * 10 ,
         n_content = sample_N_ug / sample_weight_mg ,
         PI_N_Uptake = (aboveground_biomass * n_content) / 1000 #n uptake in kg per ha
         ) #processes the data
gs_ndvi_data <- gs_ndvi_data %>%
 rowwise() %>%
 mutate(NDVI = mean(c( NDVI_1 , NDVI_2 , NDVI_3 , NDVI_4) , na.rm = T)) #takes average of four NDVI re
gs_ndvi_data <- dplyr::select(gs_ndvi_data ,</pre>
                     site_year,
                    year,
                     exp_plot_number,
                    block,
                    plot,
                    N_level_kgha,
                    aboveground_biomass,
```

```
n_content,
PI_N_Uptake,
NDVI) #selects the relevant columns

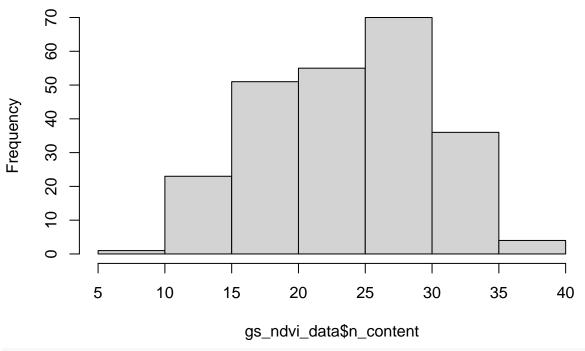
gs_ndvi_data$site_year <- factor(gs_ndvi_data$site_year , levels = c( "Nicolaus-17" , "Williams-17" ,
hist(gs_ndvi_data$aboveground_biomass)</pre>
```

Histogram of gs_ndvi_data\$aboveground_biomass



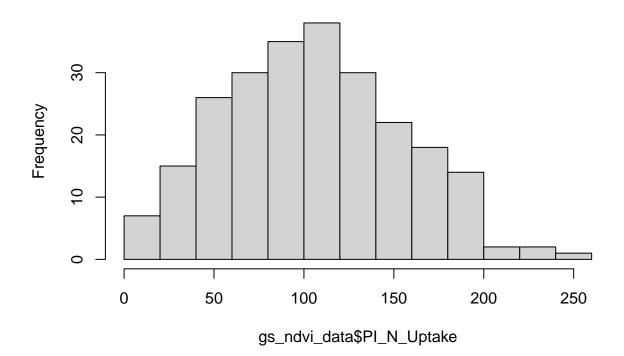
hist(gs_ndvi_data\$n_content)

Histogram of gs_ndvi_data\$n_content



hist(gs_ndvi_data\$PI_N_Uptake)

Histogram of gs_ndvi_data\$PI_N_Uptake

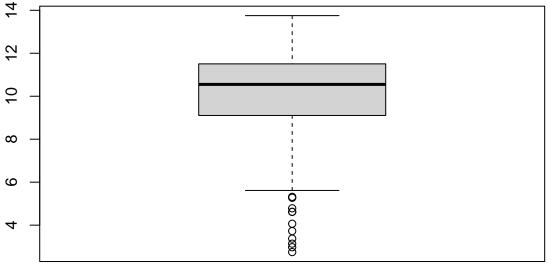


Yield Data

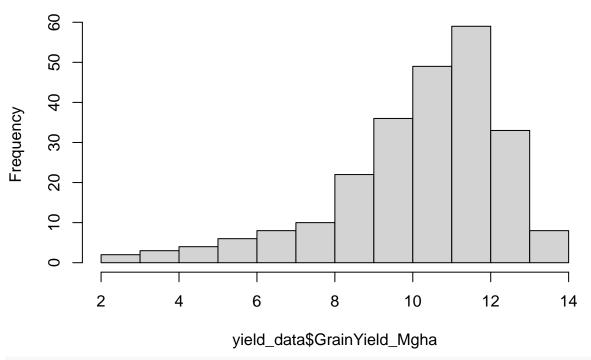
```
yield_data <- read_csv(file = "DATA/yield_data.csv")</pre>
## Rows: 672 Columns: 34
## -- Column specification -----
## Delimiter: ","
## chr (2): site_year, Received_TopDress
## dbl (32): year, Block, MainPlot, SubPlot, exp_plot_number, N_level, TopDress...
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
str(yield_data , give.attr = FALSE)
## spec_tbl_df [672 x 34] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
                      : chr [1:672] "Davis-16" "Davis-16" "Davis-16" "Davis-16" ...
## $ site_year
## $ year
                      : num [1:672] 2016 2016 2016 2016 2016 ...
## $ Block
                      : num [1:672] 1 1 1 1 1 2 2 2 2 2 ...
                      : num [1:672] 1 2 3 4 5 1 2 3 4 5 ...
## $ MainPlot
## $ SubPlot
                      : num [1:672] 3 1 3 2 1 2 3 2 3 2 ...
                    : num [1:672] 101 102 103 104 105 201 202 203 204 205 ...
## $ exp_plot_number
## $ N_level
                      : num [1:672] 75 225 0 125 175 75 0 225 175 125 ...
## $ TopDress
                      : num [1:672] 0 0 0 0 0 0 0 0 0 0 ...
## $ SeasonalNRate
                      : num [1:672] 75 225 0 125 175 75 0 225 175 125 ...
## $ N level kgha
                      : num [1:672] 75 225 0 125 175 75 0 225 175 125 ...
                      : num [1:672] 0 0 0 0 0 0 0 0 0 ...
## $ TopDress_kgha
## $ SeasonalNRate_kgha : num [1:672] 75 225 0 125 175 75 0 225 175 125 ...
## $ Received_TopDress
                      : chr [1:672] "NO" "NO" "NO" "NO" ...
## $ tare
                      ## $ FW1PlusTare
                      : num [1:672] 5662 5298 5256 4846 5220 ...
## $ FW2PlusTare
                      : num [1:672] 4610 5438 1220 4278 5156 ...
## $ SSFWPlusTare
                      : num [1:672] 3070 3262 2266 2482 2650 ...
                      : num [1:672] 692 705 519 541 538 ...
## $ SSODW
                      : num [1:672] 121 122 122 121 122 ...
## $ HarvestBagPlusTie
## $ Grain1PlusPaperBag1 : num [1:672] 292 271 188 215 208 ...
## $ PaperBag1
                      ## $ PaperBag2
## $ StrawSampleSize
                      : num [1:672] 8.27 7.74 7.9 8.38 7.52 ...
## $ StrawN
                      : num [1:672] 55.1 55.2 49.4 47.1 45.3 ...
## $ GrainSampleSize
                      : num [1:672] 4.95 4.97 5.04 5.02 4.57 ...
## $ GrainN
                      : num [1:672] 54.5 58.6 47.4 52.5 46.1 ...
## $ DaysPI2Harvest
                      ## $ SeedTray1.1
                      : num [1:672] 243 243 243 243 243 243 243 243 243 ...
## $ SeedTray1.2
                      : num [1:672] 0.62 0.62 0.62 0.64 0.64 0.62 0.62 0.62 0.62 0.62 ...
## $ Grain3PlusSeedTray1 : num [1:672] 435 514 426 455 447 ...
## $ SeedTray2
                      : num [1:672] 244 244 244 244 ...
## $ Grain4PlusSeedTray2 : num [1:672] 254 254 254 254 254 ...
yield_data <- yield_data %>%
 filter(!year %in% c( "2016") & TopDress_kgha == 0 & N_level_kgha != 275) #removing the years and N r
yield_data <- yield_data %>%
```

```
mutate(
   site_year = factor(site_year),
   year = factor(year),
   Block = factor(Block),
   MainPlot = factor(MainPlot),
   exp_plot_number = factor(exp_plot_number),
   N_level = factor(N_level),
   SubPlot = factor(SubPlot),
   TopDress = factor(TopDress),
   SeasonalNRate_f = factor(SeasonalNRate),
   N_level_kgha_f = factor(N_level_kgha),
   TopDress_kgha_f = factor(TopDress_kgha),
   SeasonalNRate_kgha_f = factor(SeasonalNRate_kgha)
   ) #changes these columns to factor
str(yield_data , give.attr = FALSE)
## tibble [240 x 38] (S3: tbl_df/tbl/data.frame)
                       : Factor w/ 10 levels "Arbuckle-18",..: 7 7 7 7 7 7 7 7 7 7 ...
## $ site_year
                       : Factor w/ 3 levels "2017", "2018", ...: 1 1 1 1 1 1 1 1 1 1 ...
## $ year
                       : Factor w/ 4 levels "1", "2", "3", "4": 1 1 1 1 1 1 2 2 2 2 ...
## $ Block
## $ MainPlot
                       : Factor w/ 7 levels "1", "2", "3", "4", ...: 1 2 3 4 6 7 2 3 4 5 ...
## $ SubPlot
                      : Factor w/ 3 levels "1", "2", "3": 2 3 1 1 1 2 1 1 2 3 ...
                      : Factor w/ 27 levels "101", "102", "103", ...: 1 2 3 4 6 7 9 10 11 12 ...
## $ exp_plot_number
                       : Factor w/ 11 levels "0", "45", "75", ...: 11 1 6 8 2 3 3 6 1 8 ...
## $ N_level
## $ TopDress
                       : Factor w/ 1 level "0": 1 1 1 1 1 1 1 1 1 ...
## $ SeasonalNRate
                       : num [1:240] 225 0 125 175 45 75 75 125 0 175 ...
## $ N_level_kgha
                       : num [1:240] 225 0 125 175 45 75 75 125 0 175 ...
## $ TopDress kgha
                       : num [1:240] 0 0 0 0 0 0 0 0 0 0 ...
## $ SeasonalNRate_kgha
                      : num [1:240] 225 0 125 175 45 75 75 125 0 175 ...
                       : chr [1:240] "NO" "NO" "NO" "NO" ...
## $ Received_TopDress
## $ tare
                       ## $ FW1PlusTare
                       : num [1:240] 4360 4818 5376 5598 4852 ...
## $ FW2PlusTare
                       : num [1:240] 3254 1220 1220 1220 1220 ...
## $ SSFWPlusTare
                       : num [1:240] 2324 1814 1994 2126 2106 ...
                       : num [1:240] 632 479 570 562 583 ...
## $ SSODW
##
   $ HarvestBagPlusTie
                       : num [1:240] 162 176 169 121 120 ...
## $ Grain1PlusPaperBag1 : num [1:240] 230 173 211 233 265 ...
                       ## $ PaperBag1
## $ PaperBag2
                       ## $ StrawSampleSize
                       : num [1:240] 8.25 7.96 7.76 7.89 8.45 ...
## $ StrawN
                       : num [1:240] 63.5 36 46.1 55.9 39.7 ...
## $ GrainSampleSize
                       : num [1:240] 4.97 5.52 5.23 5.72 5.82 ...
## $ GrainN
                       : num [1:240] 60.5 50 56 66.7 51.1 ...
## $ DaysPI2Harvest
                       : num [1:240] 76 76 76 76 76 76 76 76 76 76 ...
                       : num [1:240] 243 243 243 243 243 243 243 243 243 ...
## $ SeedTray1.1
## $ SeedTray1.2
                       ## $ Grain3PlusSeedTray1 : num [1:240] 470 410 449 471 504 ...
## $ SeedTray2
                       : num [1:240] 244 244 244 244 ...
## $ Grain4PlusSeedTray2 : num [1:240] 254 254 254 254 254 ...
                      : Factor w/ 11 levels "0", "45", "75", ...: 11 1 6 8 2 3 3 6 1 8 ...
## $ SeasonalNRate_f
## $ N_level_kgha_f
                      : Factor w/ 11 levels "0","45","75",..: 10 1 5 8 2 3 3 5 1 8 ...
```

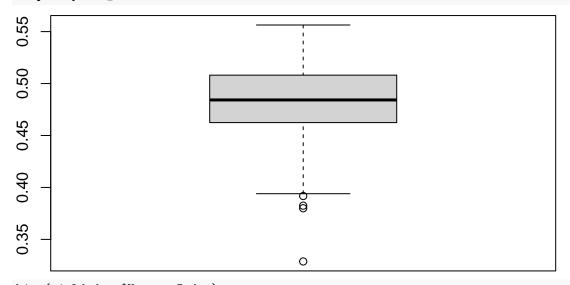
```
## $ TopDress kgha f
                       : Factor w/ 1 level "0": 1 1 1 1 1 1 1 1 1 ...
## $ SeasonalNRate_kgha_f : Factor w/ 11 levels "0","45","75",...: 10 1 5 8 2 3 3 5 1 8 ...
yield_data <- yield_data %>%
  mutate(
    FW1net = FW1PlusTare - tare,
    FW2net = FW2PlusTare - tare,
    TotalFW = FW1net + FW2net,
    SSFWnet = SSFWPlusTare - tare,
    Ratio = SSFWnet / TotalFW,
    SSODWnet = SSODW - HarvestBagPlusTie,
    SeedTray1 = SeedTray1.1 + SeedTray1.2, #adds the decimal to the 243g to get the tare weight for the
    Grain3net = Grain3PlusSeedTray1 - SeedTray1, #subtract tare of seed tray from grain3. Grain3 is the
    Grain4net = Grain4PlusSeedTray2 - SeedTray2, #grain4 is the amount of grain removed for ballmilling
    Grain2net = Grain2PlusPaperBag2 - PaperBag2, #yield component grain sample
    Grain2net = Grain2net * Ratio, #this essentially subsamples the yield component grain sample
    GrainNet = Grain3net + Grain4net + Grain2net, #add the grain removed for ball milling and yield com
    GrainRing = GrainNet / Ratio, #the amount of grain in the entire m^2 ring in grams
    GrainYield = GrainRing * 10, #g/m^2 to kg/ha
    GrainYield_kgha = GrainYield * ((100-MoistureContentGrain3)/86), #corrects for 14% moisture based on
    GrainYield_Mgha = GrainYield_kgha / 1000 , #converts kg/ha to Mg/ha
    Grain5 = GrainRing * ((100-MoistureContentGrain3)/98.1), #grain in the ring if the subsample was at
    Grain6 = GrainNet * ((100-MoistureContentGrain3)/98.1), #grain in the subsample if it was at 1.9% m
    StrawSS = SSODWnet - Grain6 , #just straw in subsample in grams
    StrawRing = StrawSS / Ratio, #straw in ring in grams i.e g/m2
    StrawNcon = StrawN / StrawSampleSize,
    StrawNup = (StrawRing * StrawNcon) / 100, #straw Nup divide by 100 to convert mg/m2 to kg/ha - this
    GrainNcon = (GrainN / GrainSampleSize), #grain in ring in kg/ha
    GrainNup = (Grain5 * GrainNcon) / 100, #grain Nup divide by 100 to convert mg N/m2 to kg N/ha
    TotalSeasonalNup = StrawNup + GrainNup, #in kg/ha
    HarvestIndex = Grain5 / (Grain5 + StrawRing),
    Moisture = SSFWnet / SSODWnet
boxplot(yield_data$GrainYield_Mgha)
```



Histogram of yield_data\$GrainYield_Mgha

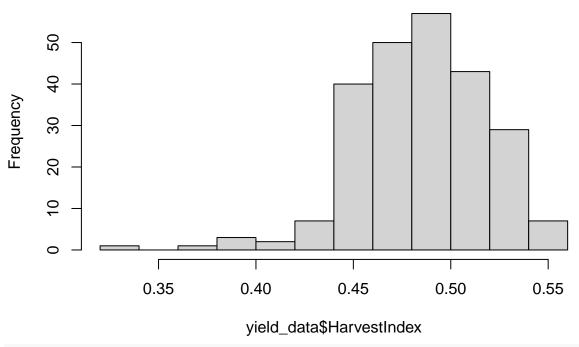


boxplot(yield_data\$HarvestIndex)

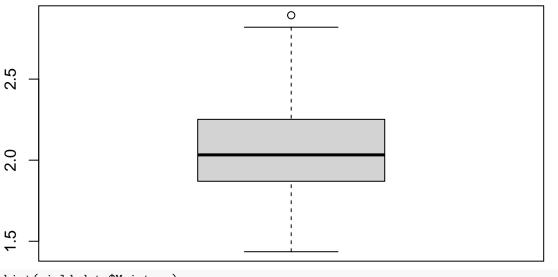


hist(yield_data\$HarvestIndex)

Histogram of yield_data\$HarvestIndex

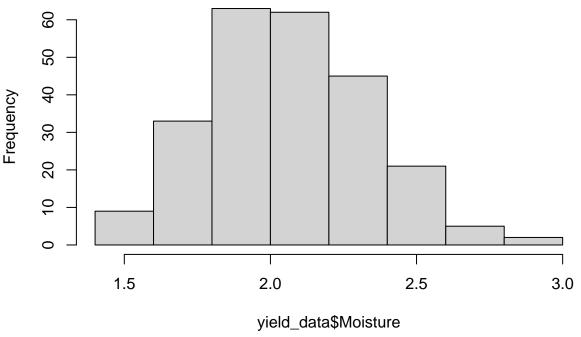


boxplot(yield_data\$Moisture)



hist(yield_data\$Moisture)

Histogram of yield_data\$Moisture

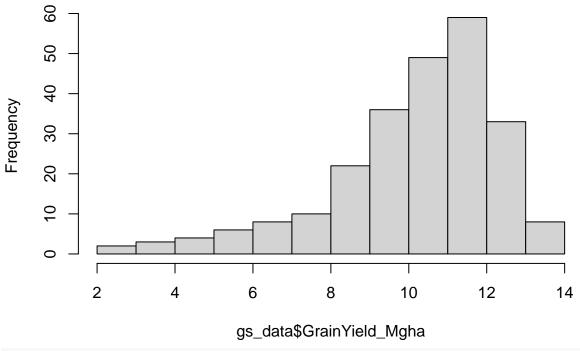


```
#the data looks good - don't see any unusual values
yield_data <- dplyr::select(yield_data,</pre>
                                 site_year,
                                 year,
                                 Block,
                                 MainPlot,
                                 SubPlot,
                                 exp_plot_number,
                                 N_level_kgha,
                                 N_level_kgha_f,
                                 TopDress_kgha,
                                 TopDress_kgha_f,
                                 GrainYield_Mgha
gs_data <- full_join(gs_ndvi_data , yield_data)</pre>
## Joining, by = c("site_year", "year", "exp_plot_number", "N_level_kgha")
gs_data <- dplyr::select(gs_data,</pre>
               site_year,
                year,
                exp_plot_number,
               Block,
               MainPlot,
               SubPlot,
                N_level_kgha,
```

```
N_level_kgha_f,
               TopDress_kgha,
               TopDress_kgha_f ,
               PI_N_Uptake,
               NDVI,
               GrainYield_Mgha) #reorders the columns
gs_data$site_year <- as.factor(gs_data$site_year)</pre>
str(gs_data , give.attr = FALSE)
## rowwise_df [240 x 13] (S3: rowwise_df/tbl_df/tbl/data.frame)
   $ site_year
                     : Factor w/ 10 levels "Nicolaus-17",..: 1 1 1 1 1 1 1 1 1 1 ...
                     : Factor w/ 3 levels "2017","2018",...: 1 1 1 1 1 1 1 1 1 1 ...
##
   $ year
## $ exp_plot_number: Factor w/ 27 levels "101","102","103",..: 1 2 3 4 6 7 9 10 11 12 ...
                    : Factor w/ 4 levels "1", "2", "3", "4": 1 1 1 1 1 2 2 2 2 ...
## $ Block
                     : Factor w/ 7 levels "1", "2", "3", "4", ...: 1 2 3 4 6 7 2 3 4 5 ...
## $ MainPlot
## $ SubPlot
                     : Factor w/ 3 levels "1", "2", "3": 2 3 1 1 1 2 1 1 2 3 ...
## $ N_level_kgha : num [1:240] 225 0 125 175 45 75 75 125 0 175 ...
## $ N_level_kgha_f : Factor w/ 11 levels "0","45","75",...: 10 1 5 8 2 3 3 5 1 8 ...
## $ TopDress_kgha : num [1:240] 0 0 0 0 0 0 0 0 0 0 ...
## $ TopDress kgha f: Factor w/ 1 level "0": 1 1 1 1 1 1 1 1 1 1 1 ...
## $ PI_N_Uptake
                    : num [1:240] 158 78 142 197 95 ...
                     : num [1:240] 0.777 0.493 0.683 0.787 0.683 ...
## $ GrainYield_Mgha: num [1:240] 11.4 11 12 11.9 11.5 ...
boxplot(gs_data$GrainYield_Mgha)
\infty
9
                                         0
```

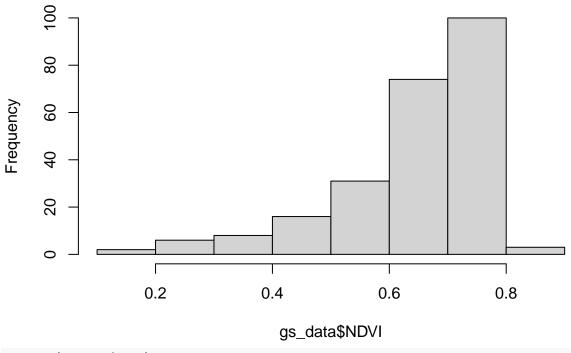
hist(gs_data\$GrainYield_Mgha)

Histogram of gs_data\$GrainYield_Mgha

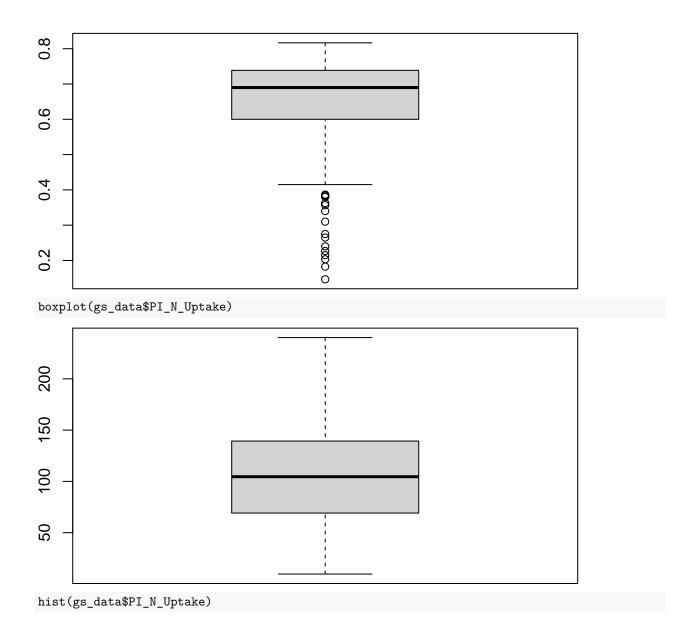


hist(gs_data\$NDVI)

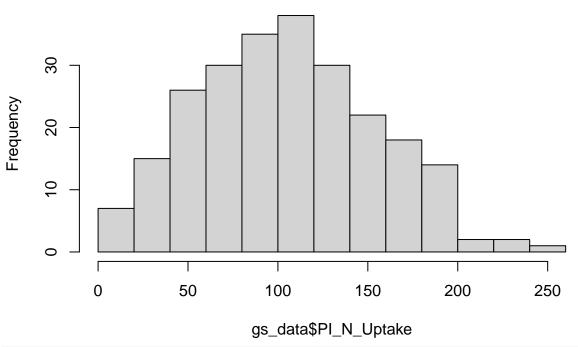
Histogram of gs_data\$NDVI



boxplot(gs_data\$NDVI)



Histogram of gs_data\$PI_N_Uptake



#Overall data looks good -- no errors of data entry

Calculating GS RI

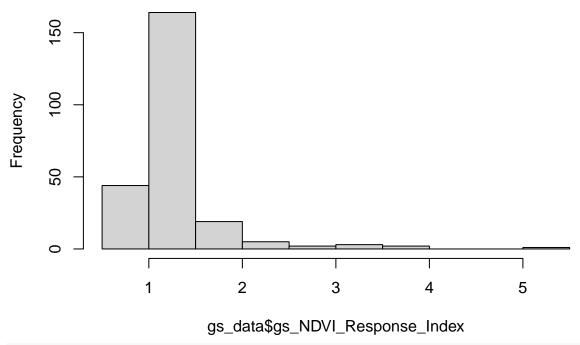
```
max_gs_ndvi <- gs_data %>%
  filter(N_level_kgha_f %in% c(225, 235)) %>%
  dplyr::select(site_year , NDVI) %>%
  group_by(site_year) %>%
  summarise(mean NDVI = mean(NDVI))
nic17 <- subset(max_gs_ndvi, site_year == "Nicolaus-17")</pre>
nic17maxNDVI <- nic17$mean_NDVI</pre>
nic17maxNDVI <- as.numeric(nic17maxNDVI)</pre>
nic17maxNDVI
## [1] 0.7858333
wil17 <- subset(max_gs_ndvi, site_year == "Williams-17")</pre>
wil17maxNDVI <- wil17$mean_NDVI</pre>
wil17maxNDVI <- as.numeric(wil17maxNDVI)</pre>
wil17maxNDVI
## [1] 0.7925
arb18 <- subset(max_gs_ndvi, site_year == "Arbuckle-18")</pre>
arb18maxNDVI <- arb18$mean_NDVI</pre>
arb18maxNDVI <- as.numeric(arb18maxNDVI)</pre>
arb18maxNDVI
```

[1] 0.73875

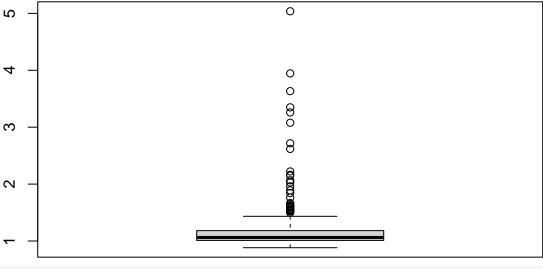
```
biggs18 <- subset(max_gs_ndvi, site_year == "Biggs-18")</pre>
biggs18maxNDVI <- biggs18$mean_NDVI</pre>
biggs18maxNDVI <- as.numeric(biggs18maxNDVI)</pre>
biggs18maxNDVI
## [1] 0.784375
mry18 <- subset(max_gs_ndvi, site_year == "Marysville-18")</pre>
mry18maxNDVI <- mry18$mean_NDVI</pre>
mry18maxNDVI <- as.numeric(mry18maxNDVI)</pre>
mry18maxNDVI
## [1] 0.728125
nic18 <- subset(max_gs_ndvi, site_year == "Nicolaus-18")</pre>
nic18maxNDVI <- nic18$mean NDVI</pre>
nic18maxNDVI <- as.numeric(nic18maxNDVI)</pre>
nic18maxNDVI
## [1] 0.765
arb19 <- subset(max_gs_ndvi, site_year == "Arbuckle-19")</pre>
arb19maxNDVI <- arb19$mean_NDVI</pre>
arb19maxNDVI <- as.numeric(arb19maxNDVI)</pre>
arb19maxNDVI
## [1] 0.72
davis19 <- subset(max_gs_ndvi, site_year == "Davis-19")</pre>
davis19maxNDVI <- davis19$mean_NDVI</pre>
davis19maxNDVI <- as.numeric(davis19maxNDVI)</pre>
davis19maxNDVI
## [1] 0.613125
mry19 <- subset(max_gs_ndvi, site_year == "Marysville-19")</pre>
mry19maxNDVI <- mry19$mean_NDVI</pre>
mry19maxNDVI <- as.numeric(mry19maxNDVI)</pre>
mry19maxNDVI
## [1] 0.745625
res19 <- subset(max_gs_ndvi, site_year == "RES-19")
res19maxNDVI <- res19$mean_NDVI</pre>
res19maxNDVI <- as.numeric(res19maxNDVI)</pre>
res19maxNDVI
## [1] 0.66625
gs_data <- gs_data %>%
  mutate(max_NDVI = case_when(
                            site_year == "Nicolaus-17" ~ nic17maxNDVI ,
                            site_year == "Williams-17" ~ wil17maxNDVI ,
                            site_year == "Arbuckle-18" ~ arb18maxNDVI ,
                            site_year == "Biggs-18" ~ biggs18maxNDVI ,
                            site_year == "Marysville-18" ~ mry18maxNDVI ,
                            site_year == "Nicolaus-18" ~ nic18maxNDVI,
                            site_year == "Arbuckle-19" ~ arb19maxNDVI ,
                            site_year == "Davis-19" ~ davis19maxNDVI ,
```

```
site_year == "Marysville-19" ~ mry19maxNDVI ,
                          site_year == "RES-19" ~ res19maxNDVI)
  )
gs_data <- gs_data %>%
  mutate(gs_NDVI_Response_Index = case_when(
                          site_year == "Nicolaus-17" ~ nic17maxNDVI / NDVI,
                          site_year == "Williams-17" ~ wil17maxNDVI / NDVI,
                          site year == "Arbuckle-18" ~ arb18maxNDVI / NDVI ,
                          site_year == "Biggs-18" ~ biggs18maxNDVI / NDVI ,
                          site_year == "Marysville-18" ~ mry18maxNDVI / NDVI ,
                          site_year == "Nicolaus-18" ~ nic18maxNDVI / NDVI,
                          site_year == "Arbuckle-19" ~ arb19maxNDVI / NDVI,
                          site_year == "Davis-19" ~ davis19maxNDVI / NDVI,
                          site_year == "Marysville-19" ~ mry19maxNDVI / NDVI,
                          site_year == "RES-19" ~ res19maxNDVI / NDVI
                        )) #calculates NDVI response index
str(gs_data , give.attr = FALSE)
## rowwise_df [240 x 15] (S3: rowwise_df/tbl_df/tbl/data.frame)
                            : Factor w/ 10 levels "Nicolaus-17",..: 1 1 1 1 1 1 1 1 1 1 ...
## $ site year
## $ year
                            : Factor w/ 3 levels "2017", "2018", ...: 1 1 1 1 1 1 1 1 1 1 ...
## $ exp_plot_number
                            : Factor w/ 27 levels "101", "102", "103", ...: 1 2 3 4 6 7 9 10 11 12 ...
## $ Block
                            : Factor w/ 4 levels "1", "2", "3", "4": 1 1 1 1 1 2 2 2 2 ...
                            : Factor w/ 7 levels "1","2","3","4",...: 1 2 3 4 6 7 2 3 4 5 ...
## $ MainPlot
## $ SubPlot
                            : Factor w/ 3 levels "1", "2", "3": 2 3 1 1 1 2 1 1 2 3 ...
## $ N level kgha
                            : num [1:240] 225 0 125 175 45 75 75 125 0 175 ...
## $ N_level_kgha_f
                            : Factor w/ 11 levels "0", "45", "75", ...: 10 1 5 8 2 3 3 5 1 8 ...
## $ TopDress_kgha
                            : num [1:240] 0 0 0 0 0 0 0 0 0 0 ...
## $ TopDress_kgha_f
                            : Factor w/ 1 level "0": 1 1 1 1 1 1 1 1 1 1 ...
## $ PI_N_Uptake
                            : num [1:240] 158 78 142 197 95 ...
## $ NDVI
                            : num [1:240] 0.777 0.493 0.683 0.787 0.683 ...
                            : num [1:240] 11.4 11 12 11.9 11.5 ...
## $ GrainYield_Mgha
                            : num [1:240] 0.786 0.786 0.786 0.786 ...
## $ max NDVI
## $ gs_NDVI_Response_Index: num [1:240] 1.012 1.593 1.15 0.999 1.15 ...
hist(gs_data$gs_NDVI_Response_Index)
```

Histogram of gs_data\$gs_NDVI_Response_Index

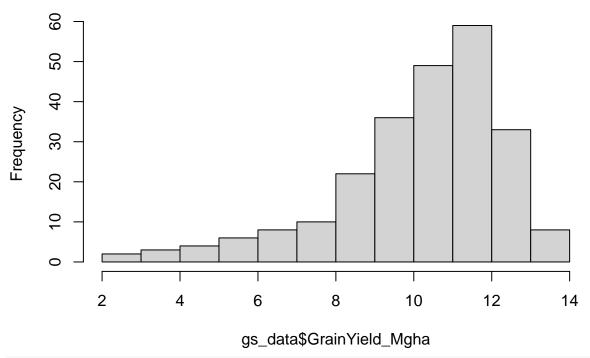


boxplot(gs_data\$gs_NDVI_Response_Index)

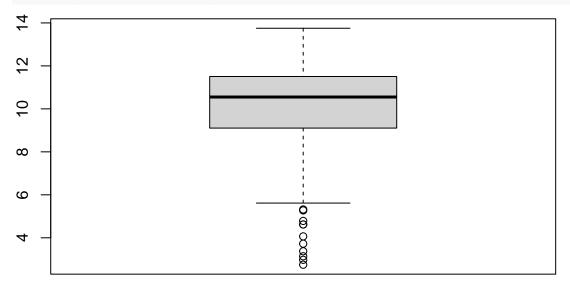


hist(gs_data\$GrainYield_Mgha)

Histogram of gs_data\$GrainYield_Mgha



boxplot(gs_data\$GrainYield_Mgha)



Drone Data

```
drone_data <- read_csv(file = "DATA/PI_drone_data.csv")

## Rows: 248 Columns: 38

## -- Column specification ------

## Delimiter: ","

## chr (1): site_year

## dbl (37): year, exp_plot_number, Block, MainPlot, N_level, N_level_kgha, bio...

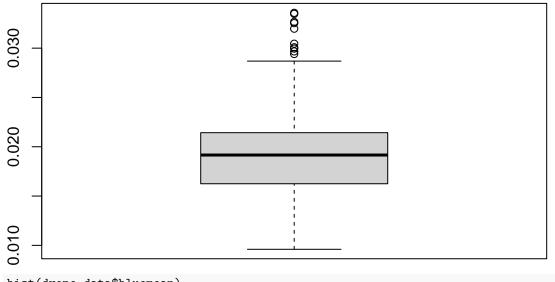
##</pre>
```

```
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
str(drone_data , give.attr = FALSE)
## spec_tbl_df [248 x 38] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
## $ site_year
                      : chr [1:248] "Nicolaus-17" "Nicolaus-17" "Nicolaus-17" "Nicolaus-17" ...
## $ year
                       : num [1:248] 2017 2017 2017 2017 ...
## $ exp_plot_number
                      : num [1:248] 101 102 103 104 105 106 107 201 202 203 ...
## $ Block
                      : num [1:248] 1 1 1 1 1 1 1 2 2 2 ...
                      : num [1:248] 1 2 3 4 5 6 7 1 2 3 ...
## $ MainPlot
## $ N level
                      : num [1:248] 225 0 125 175 275 45 75 275 75 125 ...
## $ N_level_kgha
                    : num [1:248] 225 0 125 175 275 45 75 275 75 125 ...
## $ biomass_plus_bag_g: num [1:248] 361 264 318 360 394 ...
## $ ring_size
                      ## $ paper_bag_g
                      ## $ num_of_paper_bags : num [1:248] 1 1 1 1 1 1 1 1 1 1 ...
## $ sample_weight_mg : num [1:248] 3.17 3.42 3.09 3.07 3.35 ...
## $ sample_N_ug
                      : num [1:248] 79.4 61.1 80.7 95.9 111.3 ...
## $ greenmean
                      : num [1:248] 0.0467 0.0581 0.0498 0.0488 0.0505 0.0608 0.0589 0.0461 0.0537 0.
## $ greenmedia
                      : num [1:248] NA ...
## $ greenstdev
                      : num [1:248] NA ...
## $ greenmin
                      : num [1:248] NA ...
## $ greenmax
                      : num [1:248] NA ...
## $ bluemean
                      : num [1:248] NA ...
                      : num [1:248] NA ...
## $ bluemedian
## $ bluestdev
                      : num [1:248] NA ...
## $ bluemin
                      : num [1:248] NA ...
                      : num [1:248] NA ...
## $ bluemax
##
   $ redmean
                      : num [1:248] 0.023 0.0286 0.0243 0.0237 0.0245 0.0292 0.0281 0.0234 0.0263 0.0
## $ redmedian
                      : num [1:248] NA ...
## $ redstdev
                      : num [1:248] NA ...
                      : num [1:248] NA ...
## $ redmin
## $ redmax
                      : num [1:248] NA ...
## $ edgemean
                      : num [1:248] 0.0738 0.0867 0.0768 0.0764 0.0791 0.0903 0.0877 0.0744 0.0812 0.
## $ edgemedian
                      : num [1:248] NA ...
                      : num [1:248] NA ...
## $ edgestdev
##
   $ edgemin
                      : num [1:248] NA ...
## $ edgemax
                      : num [1:248] NA ...
## $ nirmean
                      : num [1:248] 0.55 0.357 0.482 0.537 0.599 ...
## $ nirmedian
                      : num [1:248] NA ...
## $ nirstdev
                      : num [1:248] NA ...
## $ nirmin
                      : num [1:248] NA ...
## $ nirmax
                       : num [1:248] NA ...
drone_data <- drone_data %>%
 filter(N level kgha != 275) %>%
 mutate(year = factor(year) ,
        exp_plot_number = factor(exp_plot_number) ,
        Block = factor(Block) ,
        MainPlot = factor(MainPlot) ,
        N_level = factor(N_level) ,
        N_level_kgha_f = factor(N_level_kgha)
 )
```

```
drone_data$site_year <- factor(drone_data$site_year , levels = c("Nicolaus-17" , "Williams-17" , "Arbuc
str(drone_data , give.attr = FALSE)
## tibble [240 x 39] (S3: tbl_df/tbl/data.frame)
   $ site_year
                       : Factor w/ 10 levels "Nicolaus-17",..: 1 1 1 1 1 1 1 1 1 1 ...
## $ year
                       : Factor w/ 3 levels "2017", "2018", ...: 1 1 1 1 1 1 1 1 1 1 ...
## $ exp_plot_number
                       : Factor w/ 27 levels "101", "102", "103", ...: 1 2 3 4 6 7 9 10 11 12 ...
                       : Factor w/ 4 levels "1", "2", "3", "4": 1 1 1 1 1 1 2 2 2 2 ...
## $ Block
                       : Factor w/ 7 levels "1","2","3","4",...: 1 2 3 4 6 7 2 3 4 5 ...
## $ MainPlot
## $ N level
                       : Factor w/ 11 levels "0", "45", "75", ...: 11 1 6 8 2 3 3 6 1 8 ...
## $ N_level_kgha
                     : num [1:240] 225 0 125 175 45 75 75 125 0 175 ...
## $ biomass_plus_bag_g: num [1:240] 361 264 318 360 285 ...
## $ ring_size
                       : num [1:240] 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 ...
## $ paper_bag_g
                       ## $ num_of_paper_bags : num [1:240] 1 1 1 1 1 1 1 1 1 1 ...
## $ sample_weight_mg : num [1:240] 3.17 3.42 3.09 3.07 3.2 ...
## $ sample_N_ug
                       : num [1:240] 79.4 61.1 80.7 95.9 63.6 ...
## $ greenmean
                       : num [1:240] 0.0467 0.0581 0.0498 0.0488 0.0608 0.0589 0.0537 0.0488 0.0598 0.
## $ greenmedia
                       : num [1:240] NA ...
                       : num [1:240] NA ...
## $ greenstdev
## $ greenmin
                       : num [1:240] NA ...
## $ greenmax
                       : num [1:240] NA ...
## $ bluemean
                       : num [1:240] NA ...
                       : num [1:240] NA ...
## $ bluemedian
## $ bluestdev
                      : num [1:240] NA ...
## $ bluemin
                       : num [1:240] NA ...
                       : num [1:240] NA ...
## $ bluemax
##
   $ redmean
                       : num [1:240] 0.023 0.0286 0.0243 0.0237 0.0292 0.0281 0.0263 0.0237 0.0293 0.0
## $ redmedian
                      : num [1:240] NA ...
## $ redstdev
                      : num [1:240] NA ...
                       : num [1:240] NA ...
## $ redmin
## $ redmax
                       : num [1:240] NA ...
## $ edgemean
                       : num [1:240] 0.0738 0.0867 0.0768 0.0764 0.0903 0.0877 0.0812 0.0754 0.0873 0.
## $ edgemedian
                       : num [1:240] NA ...
                       : num [1:240] NA ...
## $ edgestdev
                       : num [1:240] NA ...
##
   $ edgemin
## $ edgemax
                       : num [1:240] NA ...
                       : num [1:240] 0.55 0.357 0.482 0.537 0.431 ...
## $ nirmean
## $ nirmedian
                       : num [1:240] NA ...
                       : num [1:240] NA ...
## $ nirstdev
## $ nirmin
                       : num [1:240] NA ...
## $ nirmax
                       : num [1:240] NA ...
                       : Factor w/ 11 levels "0", "45", "75", ...: 10 1 5 8 2 3 3 5 1 8 ...
   $ N level kgha f
drone_data <- dplyr::select(drone_data ,</pre>
                     site_year,
                     year,
                     exp_plot_number,
                     Block,
                     MainPlot,
                     N level,
                     N_level_kgha,
                     N_level_kgha_f,
```

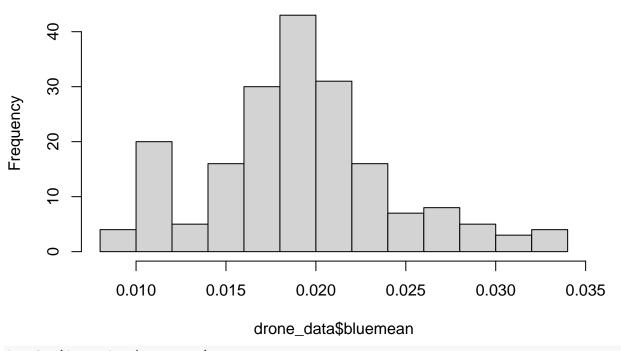
```
biomass_plus_bag_g,
                    ring_size,
                    paper_bag_g,
                    num_of_paper_bags,
                    sample_weight_mg,
                    sample_N_ug,
                    bluemean,
                    greenmean,
                    redmean,
                    edgemean,
                    nirmean
                    )#selects the relevant columns
str(drone_data , give.attr = FALSE)
## tibble [240 x 19] (S3: tbl_df/tbl/data.frame)
## $ site_year
                     : Factor w/ 10 levels "Nicolaus-17",..: 1 1 1 1 1 1 1 1 1 1 ...
                     : Factor w/ 3 levels "2017","2018",...: 1 1 1 1 1 1 1 1 1 1 ...
## $ year
## $ exp_plot_number : Factor w/ 27 levels "101","102","103",..: 1 2 3 4 6 7 9 10 11 12 ...
                     : Factor w/ 4 levels "1", "2", "3", "4": 1 1 1 1 1 2 2 2 2 ...
## $ Block
                     : Factor w/ 7 levels "1","2","3","4",...: 1 2 3 4 6 7 2 3 4 5 ...
## $ MainPlot
                     : Factor w/ 11 levels "0", "45", "75", ...: 11 1 6 8 2 3 3 6 1 8 ....
## $ N level
## $ N_level_kgha
                    : num [1:240] 225 0 125 175 45 75 75 125 0 175 ...
## $ N_level_kgha_f : Factor w/ 11 levels "0","45","75",..: 10 1 5 8 2 3 3 5 1 8 ...
## $ biomass_plus_bag_g: num [1:240] 361 264 318 360 285 ...
## $ ring_size
                    ## $ paper_bag_g
                     ## $ num_of_paper_bags : num [1:240] 1 1 1 1 1 1 1 1 1 1 ...
## $ sample_weight_mg : num [1:240] 3.17 3.42 3.09 3.07 3.2 ...
                     : num [1:240] 79.4 61.1 80.7 95.9 63.6 ...
## $ sample_N_ug
## $ bluemean
                     : num [1:240] NA ...
                     : num [1:240] 0.0467 0.0581 0.0498 0.0488 0.0608 0.0589 0.0537 0.0488 0.0598 0.
## $ greenmean
                     : num [1:240] 0.023 0.0286 0.0243 0.0237 0.0292 0.0281 0.0263 0.0237 0.0293 0.0
## $ redmean
   $ edgemean
                     : num [1:240] 0.0738 0.0867 0.0768 0.0764 0.0903 0.0877 0.0812 0.0754 0.0873 0.
                     : num [1:240] 0.55 0.357 0.482 0.537 0.431 ...
## $ nirmean
#visualize drone_data to look for outliers
```

boxplot(drone_data\$bluemean)

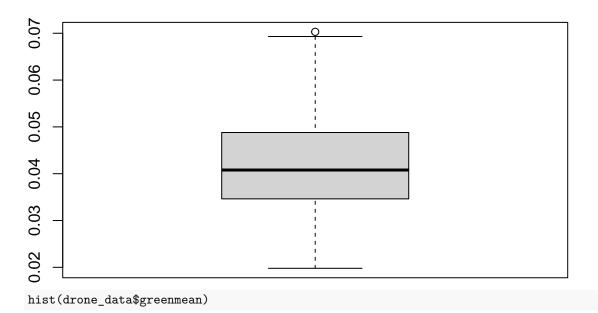


hist(drone_data\$bluemean)

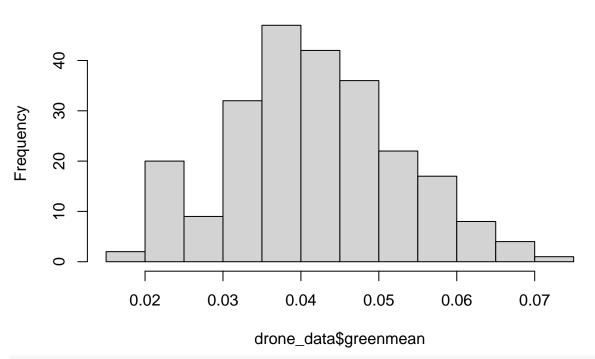
Histogram of drone_data\$bluemean



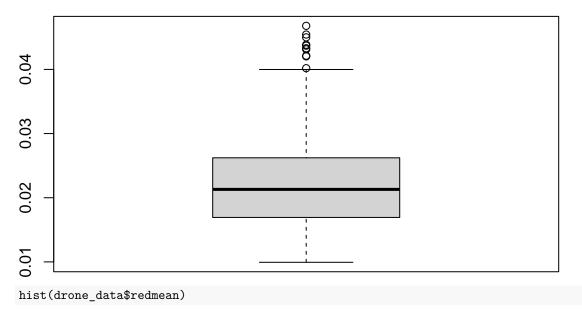
boxplot(drone_data\$greenmean)



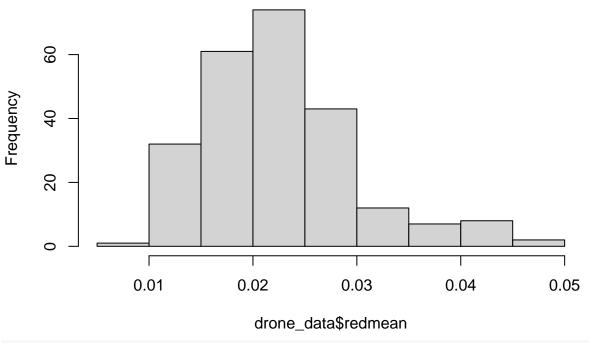
Histogram of drone_data\$greenmean



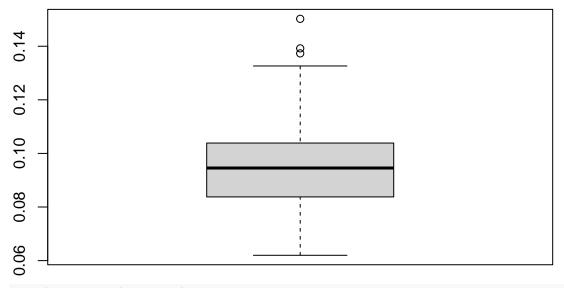
boxplot(drone_data\$redmean)



Histogram of drone_data\$redmean

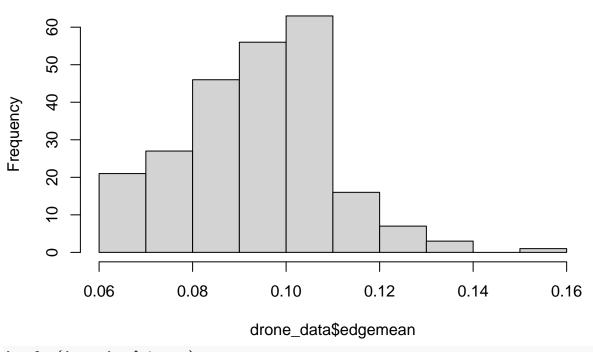


boxplot(drone_data\$edgemean)

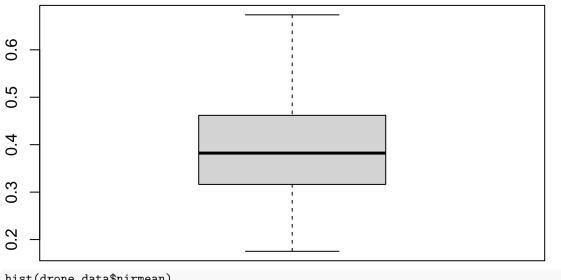


hist(drone_data\$edgemean)

Histogram of drone_data\$edgemean

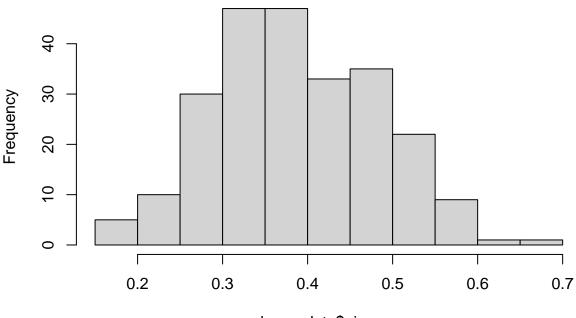


boxplot(drone_data\$nirmean)



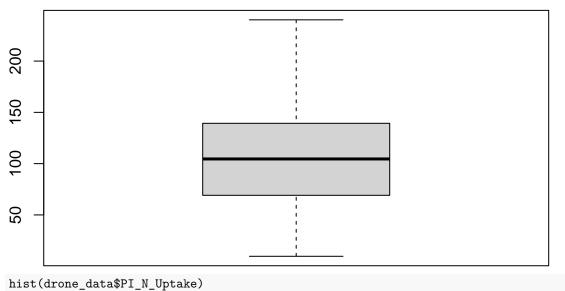
hist(drone_data\$nirmean)

Histogram of drone_data\$nirmean



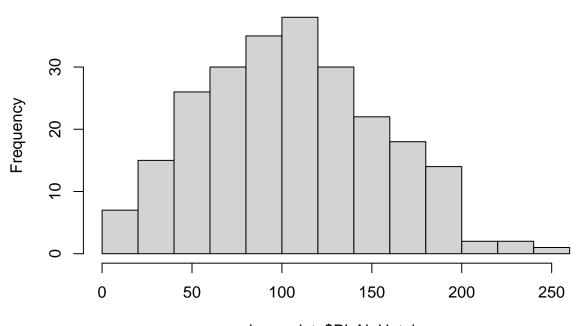
drone_data\$nirmean

```
drone_data <- drone_data %>%
  mutate( biomass_dry_wt = biomass_plus_bag_g - (paper_bag_g * num_of_paper_bags) ,
          aboveground_biomass = (biomass_dry_wt / ring_size) * 10 , #ring size 0.5 m^2 biomass in kg pe
          n_content = sample_N_ug / sample_weight_mg ,
          PI_N_Uptake = (aboveground_biomass * n_content) / 1000 #n uptake in kg per ha
          )#processes the data2
boxplot(drone_data$PI_N_Uptake)
```

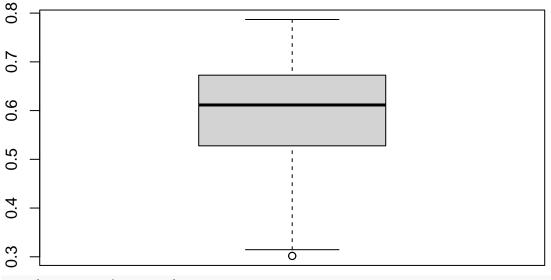


_ •

Histogram of drone_data\$PI_N_Uptake

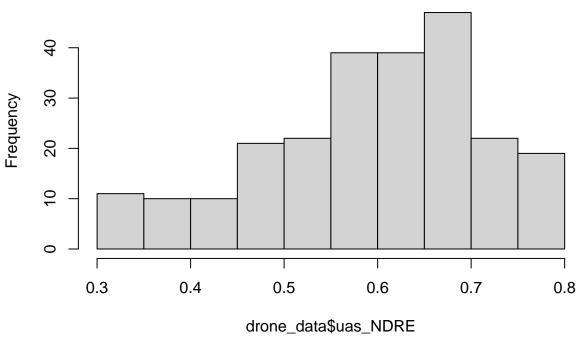


drone_data\$PI_N_Uptake

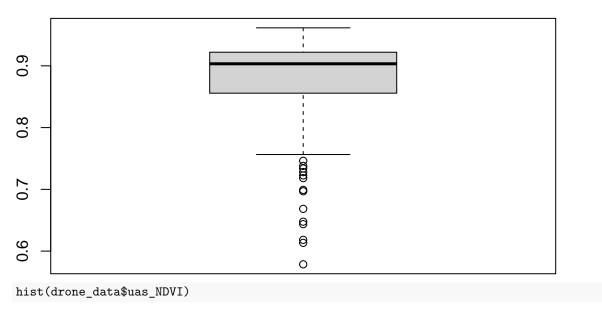


hist(drone_data\$uas_NDRE)

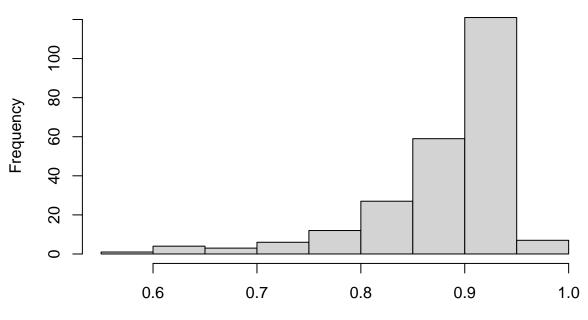
Histogram of drone_data\$uas_NDRE



boxplot(drone_data\$uas_NDVI)

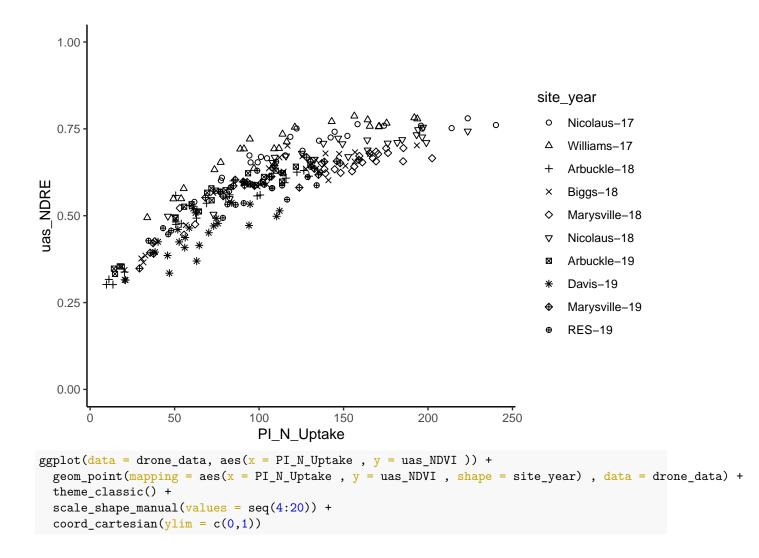


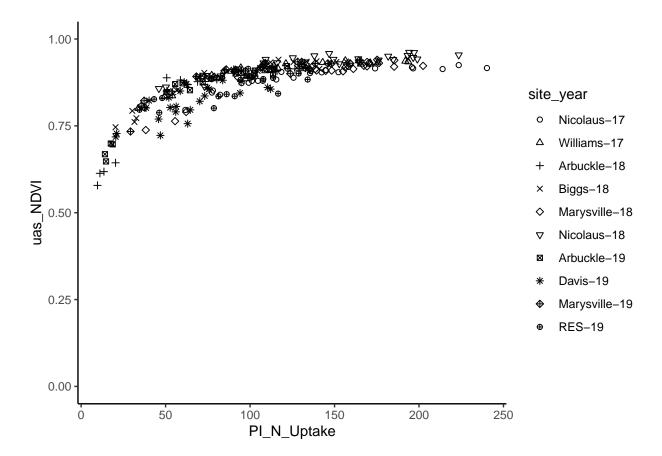
Histogram of drone_data\$uas_NDVI



drone_data\$uas_NDVI

```
ggplot(data = drone_data, aes(x = PI_N_Uptake , y = uas_NDRE )) +
geom_point(mapping = aes(x = PI_N_Uptake , y = uas_NDRE , shape = site_year) , data = drone_data) +
theme_classic() +
scale_shape_manual(values = seq(4:20)) +
coord_cartesian(ylim = c(0,1))
```





Calculating UAS RI

```
#gets the max NDRE value for each site
max_drone_data <- drone_data %>%
  filter(N_level_kgha_f %in% c(225, 235)) %>%
  dplyr::select(site_year , uas_NDVI , uas_NDRE) %>%
  group_by(site_year) %>%
  summarise_all(.funs = mean) %>%
  ungroup()
nic17 <- subset(max_drone_data, site_year == "Nicolaus-17")</pre>
nic17maxuas_NDRE <- as.numeric(nic17$uas_NDRE)</pre>
nic17maxuas_NDRE
## [1] 0.7642401
nic17maxuas_NDVI <- as.numeric(nic17$uas_NDVI)</pre>
nic17maxuas_NDVI
## [1] 0.9186193
wil17 <- subset(max_drone_data, site_year == "Williams-17")</pre>
wil17maxuas_NDRE <- as.numeric(wil17$uas_NDRE)</pre>
wil17maxuas_NDRE
```

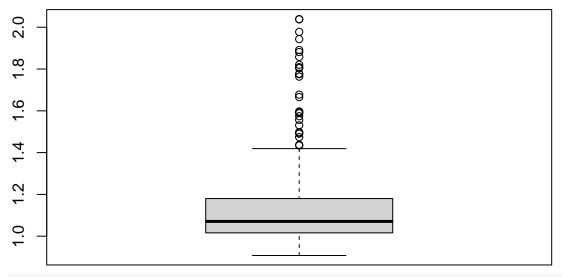
```
wil17maxuas_NDVI <- as.numeric(wil17$uas_NDVI)</pre>
wil17maxuas_NDVI
## [1] 0.9349728
arb18 <- subset(max_drone_data, site_year == "Arbuckle-18")</pre>
arb18maxuas_NDRE <- as.numeric(arb18$uas_NDRE)</pre>
arb18maxuas_NDRE
## [1] 0.6151996
arb18maxuas_NDVI <- as.numeric(arb18$uas_NDVI)</pre>
arb18maxuas_NDVI
## [1] 0.9113451
biggs18 <- subset(max_drone_data, site_year == "Biggs-18")</pre>
biggs18maxuas_NDRE <- as.numeric(biggs18$uas_NDRE)</pre>
biggs18maxuas_NDRE
## [1] 0.6796539
biggs18maxuas_NDVI <- as.numeric(biggs18$uas_NDVI)</pre>
biggs18maxuas_NDVI
## [1] 0.9342827
mry18 <- subset(max_drone_data, site_year == "Marysville-18")</pre>
mry18maxuas_NDRE <- as.numeric(mry18$uas_NDRE)</pre>
mry18maxuas_NDRE
## [1] 0.6812887
mry18maxuas_NDVI <- as.numeric(mry18$uas_NDVI)</pre>
mry18maxuas_NDVI
## [1] 0.9315267
nic18 <- subset(max_drone_data, site_year == "Nicolaus-18")</pre>
nic18maxuas_NDRE <- as.numeric(nic18$uas_NDRE)</pre>
nic18maxuas_NDRE
## [1] 0.7430412
nic18maxuas_NDVI <- as.numeric(nic18$uas_NDVI)</pre>
nic18maxuas_NDVI
## [1] 0.9558061
arb19 <- subset(max_drone_data, site_year == "Arbuckle-19")</pre>
arb19maxuas_NDRE <- as.numeric(arb19$uas_NDRE)</pre>
arb19maxuas_NDRE
## [1] 0.6288361
arb19maxuas_NDVI <- as.numeric(arb19$uas_NDVI)</pre>
arb19maxuas_NDVI
## [1] 0.9142775
```

```
davis19 <- subset(max_drone_data, site_year == "Davis-19")</pre>
davis19maxuas_NDRE <- as.numeric(davis19$uas_NDRE)</pre>
davis19maxuas_NDRE
## [1] 0.5012478
davis19maxuas_NDVI <- as.numeric(davis19$uas_NDVI)</pre>
davis19maxuas_NDVI
## [1] 0.8592071
mry19 <- subset(max_drone_data, site_year == "Marysville-19")</pre>
mry19maxuas_NDRE <- as.numeric(mry19$uas_NDRE)</pre>
mry19maxuas_NDRE
## [1] 0.655337
mry19maxuas_NDVI <- as.numeric(mry19$uas_NDVI)</pre>
mry19maxuas NDVI
## [1] 0.9276236
res19 <- subset(max_drone_data, site_year == "RES-19")
res19maxuas_NDRE <- as.numeric(res19$uas_NDRE)</pre>
res19maxuas_NDRE
## [1] 0.5945887
res19maxuas_NDVI <- as.numeric(res19$uas_NDVI)</pre>
res19maxuas_NDVI
## [1] 0.8840389
drone_data <- drone_data %>%
  mutate(max_uas_NDRE = case_when(
                           site_year == "Nicolaus-17" ~ nic17maxuas_NDRE ,
                           site_year == "Williams-17" ~ will7maxuas_NDRE ,
                           site_year == "Arbuckle-18" ~ arb18maxuas_NDRE ,
                           site_year == "Biggs-18" ~ biggs18maxuas_NDRE ,
                           site_year == "Marysville-18" ~ mry18maxuas_NDRE ,
                           site_year == "Nicolaus-18" ~ nic18maxuas_NDRE,
                           site_year == "Arbuckle-19" ~ arb19maxuas_NDRE ,
                           site_year == "Davis-19" ~ davis19maxuas_NDRE ,
                           site_year == "Marysville-19" ~ mry19maxuas_NDRE ,
                           site_year == "RES-19" ~ res19maxuas_NDRE) #assign the max NDRE value for each
  )
drone_data <- drone_data %>%
  mutate(uas_NDRE_Response_Index = case_when()
                           site_year == "Nicolaus-17" ~ nic17maxuas_NDRE / uas_NDRE,
                           site_year == "Williams-17" ~ will7maxuas_NDRE / uas_NDRE,
                           site_year == "Arbuckle-18" ~ arb18maxuas_NDRE / uas_NDRE ,
                           site_year == "Biggs-18" ~ biggs18maxuas_NDRE / uas_NDRE ,
                           site_year == "Marysville-18" ~ mry18maxuas_NDRE / uas_NDRE ,
                           site_year == "Nicolaus-18" ~ nic18maxuas_NDRE / uas_NDRE,
                           site_year == "Arbuckle-19" ~ arb19maxuas_NDRE / uas_NDRE,
                           site_year == "Davis-19" ~ davis19maxuas_NDRE / uas_NDRE,
                           site_year == "Marysville-19" ~ mry19maxuas_NDRE / uas_NDRE,
```

```
site_year == "RES-19" ~ res19maxuas_NDRE / uas_NDRE
                       )) #calculates uas_NDRE response index
drone_data <- drone_data %>%
 mutate(max_uas_NDVI = case_when(
                         site_year == "Nicolaus-17" ~ nic17maxuas_NDVI ,
                         site_year == "Williams-17" ~ wil17maxuas_NDVI ,
                         site year == "Arbuckle-18" ~ arb18maxuas NDVI ,
                         site_year == "Biggs-18" ~ biggs18maxuas_NDVI ,
                         site_year == "Marysville-18" ~ mry18maxuas_NDVI ,
                         site_year == "Nicolaus-18" ~ nic18maxuas_NDVI,
                         site_year == "Arbuckle-19" ~ arb19maxuas_NDVI ,
                         site_year == "Davis-19" ~ davis19maxuas_NDVI ,
                         site_year == "Marysville-19" ~ mry19maxuas_NDVI ,
                         site_year == "RES-19" ~ res19maxuas_NDVI) #assign max ndvi value for each sit
 )
drone_data <- drone_data %>%
 mutate(uas_NDVI_Response_Index = case_when()
                         site_year == "Nicolaus-17" ~ nic17maxuas_NDVI / uas_NDVI,
                         site_year == "Williams-17" ~ will7maxuas_NDVI / uas_NDVI,
                         site_year == "Arbuckle-18" ~ arb18maxuas_NDVI / uas_NDVI ,
                         site_year == "Biggs-18" ~ biggs18maxuas_NDVI / uas_NDVI ,
                         site_year == "Marysville-18" ~ mry18maxuas_NDVI / uas_NDVI ,
                         site_year == "Nicolaus-18" ~ nic18maxuas_NDVI / uas_NDVI,
                         site_year == "Arbuckle-19" ~ arb19maxuas_NDVI / uas_NDVI,
                         site_year == "Davis-19" ~ davis19maxuas_NDVI / uas_NDVI,
                         site_year == "Marysville-19" ~ mry19maxuas_NDVI / uas_NDVI,
                         site_year == "RES-19" ~ res19maxuas_NDVI / uas_NDVI
                       )) #calculates uas_NDVI response index
str(drone_data , give.attr = F)
## tibble [240 x 29] (S3: tbl_df/tbl/data.frame)
## $ site_year
                            : Factor w/ 10 levels "Nicolaus-17",..: 1 1 1 1 1 1 1 1 1 1 ...
## $ year
                            : Factor w/ 3 levels "2017","2018",...: 1 1 1 1 1 1 1 1 1 1 ...
                           : Factor w/ 27 levels "101", "102", "103", ...: 1 2 3 4 6 7 9 10 11 12 ....
## $ exp_plot_number
## $ Block
                           : Factor w/ 4 levels "1", "2", "3", "4": 1 1 1 1 1 1 2 2 2 2 ...
                           : Factor w/ 7 levels "1", "2", "3", "4", ...: 1 2 3 4 6 7 2 3 4 5 ...
## $ MainPlot
                           : Factor w/ 11 levels "0","45","75",..: 11 1 6 8 2 3 3 6 1 8 ...
## $ N_level
## $ N_level_kgha
                           : num [1:240] 225 0 125 175 45 75 75 125 0 175 ...
                          : Factor w/ 11 levels "0", "45", "75", ...: 10 1 5 8 2 3 3 5 1 8 ...
## $ N_level_kgha_f
## $ biomass_plus_bag_g
                          : num [1:240] 361 264 318 360 285 ...
## $ ring_size
                           ## $ paper_bag_g
                           ## $ num_of_paper_bags
                           : num [1:240] 1 1 1 1 1 1 1 1 1 1 ...
                           : num [1:240] 3.17 3.42 3.09 3.07 3.2 ...
## $ sample_weight_mg
## $ sample_N_ug
                           : num [1:240] 79.4 61.1 80.7 95.9 63.6 ...
## $ bluemean
                           : num [1:240] NA ...
## $ greenmean
                           : num [1:240] 0.0467 0.0581 0.0498 0.0488 0.0608 0.0589 0.0537 0.0488 0.05
                           : num [1:240] 0.023 0.0286 0.0243 0.0237 0.0292 0.0281 0.0263 0.0237 0.029
## $ redmean
                           : num [1:240] 0.0738 0.0867 0.0768 0.0764 0.0903 0.0877 0.0812 0.0754 0.08
## $ edgemean
                          : num [1:240] 0.55 0.357 0.482 0.537 0.431 ...
## $ nirmean
## $ biomass_dry_wt
                          : num [1:240] 316 219 273 315 239 ...
```

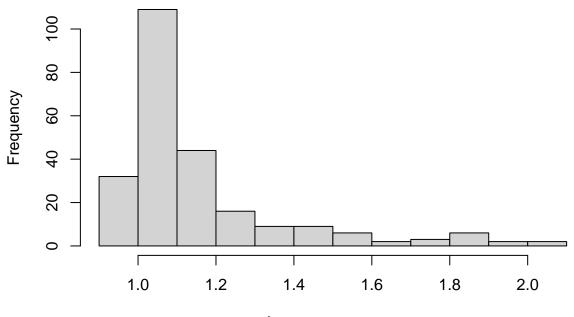
```
$ aboveground_biomass
                            : num [1:240] 6316 4376 5456 6296 4782 ...
##
   $ n_content
                            : num [1:240] 25.1 17.8 26.1 31.2 19.9 ...
   $ PI_N_Uptake
                              num [1:240] 158 78 142 197 95 ...
   $ uas_NDRE
                              num [1:240] 0.763 0.609 0.725 0.751 0.653 ...
##
##
   $ uas_NDVI
                              num [1:240] 0.92 0.852 0.904 0.915 0.873 ...
   $ max_uas_NDRE
                             : num [1:240] 0.764 0.764 0.764 0.764 ...
##
   $ uas_NDRE_Response_Index: num [1:240] 1 1.25 1.05 1.02 1.17 ...
                            : num [1:240] 0.919 0.919 0.919 0.919 ...
   $ max_uas_NDVI
   $ uas_NDVI_Response_Index: num [1:240] 0.999 1.079 1.016 1.003 1.052 ...
```

boxplot(drone_data\$uas_NDRE_Response_Index)

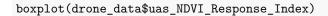


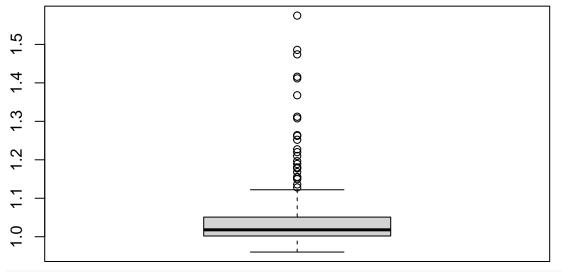
hist(drone_data\$uas_NDRE_Response_Index)

Histogram of drone_data\$uas_NDRE_Response_Index



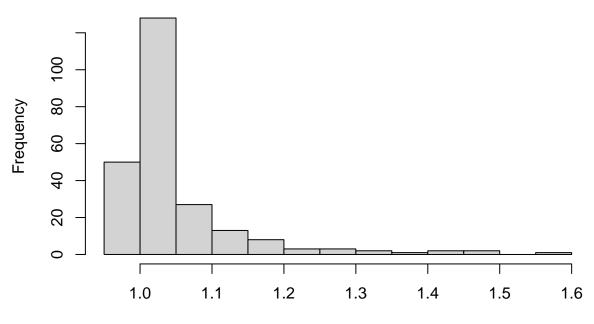
drone_data\$uas_NDRE_Response_Index





hist(drone_data\$uas_NDVI_Response_Index)

Histogram of drone_data\$uas_NDVI_Response_Index



drone_data\$uas_NDVI_Response_Index

```
uas_NDRE,
                      uas_NDVI,
                      uas NDRE Response Index,
                      uas NDVI Response Index
                      )#selects the relevant columns
sUAS_yield_data <- yield_data
sUAS_yield_data$site_year <- factor(sUAS_yield_data$site_year , levels = c("Nicolaus-17" , "Williams-17"
str(drone_data , give.attr = F)
## tibble [240 x 12] (S3: tbl_df/tbl/data.frame)
                             : Factor w/ 10 levels "Nicolaus-17",...: 1 1 1 1 1 1 1 1 1 1 ...
## $ site_year
                             : Factor w/ 3 levels "2017", "2018", ...: 1 1 1 1 1 1 1 1 1 1 ...
## $ year
## $ exp_plot_number
                            : Factor w/ 27 levels "101", "102", "103", ...: 1 2 3 4 6 7 9 10 11 12 ...
                             : Factor w/ 4 levels "1", "2", "3", "4": 1 1 1 1 1 1 2 2 2 2 ...
## $ Block
## $ MainPlot
                             : Factor w/ 7 levels "1", "2", "3", "4", ...: 1 2 3 4 6 7 2 3 4 5 ...
## $ N_level_kgha
                            : num [1:240] 225 0 125 175 45 75 75 125 0 175 ...
                            : Factor w/ 11 levels "0", "45", "75", ...: 10 1 5 8 2 3 3 5 1 8 ...
## $ N_level_kgha_f
## $ PI N Uptake
                             : num [1:240] 158 78 142 197 95 ...
## $ uas_NDRE
                             : num [1:240] 0.763 0.609 0.725 0.751 0.653 ...
## $ uas NDVI
                             : num [1:240] 0.92 0.852 0.904 0.915 0.873 ...
## $ uas NDRE Response Index: num [1:240] 1 1.25 1.05 1.02 1.17 ...
## $ uas_NDVI_Response_Index: num [1:240] 0.999 1.079 1.016 1.003 1.052 ...
str(sUAS_yield_data , give.attr = F)
## tibble [240 x 11] (S3: tbl_df/tbl/data.frame)
## $ site year
                    : Factor w/ 10 levels "Nicolaus-17",..: 1 1 1 1 1 1 1 1 1 1 ...
## $ year
                     : Factor w/ 3 levels "2017", "2018", ...: 1 1 1 1 1 1 1 1 1 1 ...
                     : Factor w/ 4 levels "1", "2", "3", "4": 1 1 1 1 1 2 2 2 2 ...
## $ Block
## $ MainPlot
                    : Factor w/ 7 levels "1", "2", "3", "4", ...: 1 2 3 4 6 7 2 3 4 5 ...
                    : Factor w/ 3 levels "1", "2", "3": 2 3 1 1 1 2 1 1 2 3 ...
## $ SubPlot
## $ exp_plot_number: Factor w/ 27 levels "101","102","103",..: 1 2 3 4 6 7 9 10 11 12 ...
## $ N_level_kgha : num [1:240] 225 0 125 175 45 75 75 125 0 175 ...
## $ N_level_kgha_f : Factor w/ 11 levels "0","45","75",...: 10 1 5 8 2 3 3 5 1 8 ...
## $ TopDress_kgha : num [1:240] 0 0 0 0 0 0 0 0 0 ...
## $ TopDress_kgha_f: Factor w/ 1 level "0": 1 1 1 1 1 1 1 1 1 1 ...
## $ GrainYield_Mgha: num [1:240] 11.4 11 12 11.9 11.5 ...
sUAS_data <- full_join( drone_data , sUAS_yield_data)</pre>
## Joining, by = c("site_year", "year", "exp_plot_number", "Block", "MainPlot",
## "N_level_kgha", "N_level_kgha_f")
sUAS_data <- dplyr::select(sUAS_data ,</pre>
                      site_year,
                      year,
                      exp_plot_number,
                      Block,
                      MainPlot,
                      SubPlot,
                      N level kgha,
                      N level kgha f,
                      TopDress_kgha,
```

```
TopDress_kgha_f,
                     PI_N_Uptake,
                     uas NDRE,
                     uas NDVI,
                     GrainYield_Mgha,
                     uas_NDRE_Response_Index,
                     uas_NDVI_Response_Index
                     )#selects the relevant columns
str(sUAS_data , give.attr = F)
## tibble [240 x 16] (S3: tbl_df/tbl/data.frame)
                            : Factor w/ 10 levels "Nicolaus-17",..: 1 1 1 1 1 1 1 1 1 1 ...
## $ site_year
                            : Factor w/ 3 levels "2017", "2018", ...: 1 1 1 1 1 1 1 1 1 1 ...
## $ year
                           : Factor w/ 27 levels "101", "102", "103", ...: 1 2 3 4 6 7 9 10 11 12 ...
## $ exp_plot_number
                            : Factor w/ 4 levels "1","2","3","4": 1 1 1 1 1 2 2 2 2 ...
## $ Block
                            : Factor w/ 7 levels "1", "2", "3", "4", ...: 1 2 3 4 6 7 2 3 4 5 ...
## $ MainPlot
## $ SubPlot
                           : Factor w/ 3 levels "1","2","3": 2 3 1 1 1 2 1 1 2 3 ...
                           : num [1:240] 225 0 125 175 45 75 75 125 0 175 ...
## $ N_level_kgha
## $ N_level_kgha_f
                           : Factor w/ 11 levels "0","45","75",...: 10 1 5 8 2 3 3 5 1 8 ...
## $ TopDress_kgha
                            : num [1:240] 0 0 0 0 0 0 0 0 0 0 ...
## $ TopDress_kgha_f
                           : Factor w/ 1 level "0": 1 1 1 1 1 1 1 1 1 ...
## $ PI N Uptake
                           : num [1:240] 158 78 142 197 95 ...
## $ uas_NDRE
                            : num [1:240] 0.763 0.609 0.725 0.751 0.653 ...
## $ uas_NDVI
                            : num [1:240] 0.92 0.852 0.904 0.915 0.873 ...
## $ GrainYield_Mgha
                           : num [1:240] 11.4 11 12 11.9 11.5 ...
## $ uas_NDRE_Response_Index: num [1:240] 1 1.25 1.05 1.02 1.17 ...
## $ uas_NDVI_Response_Index: num [1:240] 0.999 1.079 1.016 1.003 1.052 ...
Combining the data
 mutate(gs_NDVI_SI = 1 / gs_NDVI_Response_Index)
```

```
gs data <- gs data %>%
sUAS_data <- sUAS_data %>%
  mutate(uas_NDVI_SI = 1 / uas_NDVI_Response_Index)
sUAS_data <- sUAS_data %>%
  mutate(uas_NDRE_SI = 1 / uas_NDRE_Response_Index)
gs_data <- gs_data %>%
  dplyr::select(site_year ,
                year,
                exp_plot_number,
                Block,
                MainPlot,
                SubPlot,
                N_level_kgha,
                N level kgha f,
                TopDress_kgha,
                TopDress_kgha_f,
                PI_N_Uptake,
```

```
NDVI,
                GrainYield_Mgha,
                gs_NDVI_Response_Index,
                gs_NDVI_SI
         )
sUAS_data <- sUAS_data %>%
  dplyr::select(site_year ,
                year,
                exp_plot_number,
                Block,
                MainPlot,
                SubPlot,
                N_level_kgha,
                N_level_kgha_f,
                TopDress_kgha,
                TopDress_kgha_f,
                uas_NDRE,
                uas_NDVI,
                uas_NDVI_Response_Index,
                uas_NDVI_SI,
                uas_NDRE_Response_Index,
                uas_NDRE_SI
         )
paper3_data <- full_join(gs_data , sUAS_data)</pre>
## Joining, by = c("site_year", "year", "exp_plot_number", "Block", "MainPlot",
## "SubPlot", "N_level_kgha", "N_level_kgha_f", "TopDress_kgha",
## "TopDress_kgha_f")
paper3_gsdata <- paper3_data %>%
  dplyr::select(site_year ,
                year ,
                exp_plot_number ,
                Block ,
                MainPlot ,
                SubPlot ,
                N_level_kgha ,
                N_level_kgha_f ,
                TopDress kgha,
                TopDress_kgha_f ,
                NDVI,
                gs_NDVI_Response_Index,
                gs_NDVI_SI ,
                GrainYield_Mgha,
                PI_N_Uptake) %>%
  mutate(Platform = "GreenSeeker NDVI" ) %>%
  rename(Index = NDVI,
         SI = gs_NDVI_SI,
         RI = gs_NDVI_Response_Index)
paper3_uas_ndvi_data <- paper3_data %>%
 dplyr::select(site_year ,
```

```
year ,
                exp_plot_number ,
                Block ,
                MainPlot,
                SubPlot ,
                N_level_kgha ,
                N_level_kgha_f ,
                TopDress_kgha ,
                TopDress_kgha_f ,
                uas_NDVI,
                uas_NDVI_Response_Index,
                uas_NDVI_SI ,
                GrainYield_Mgha,
                PI_N_Uptake) %>%
  mutate(Platform = "sUAS NDVI" ) %>%
  rename(Index = uas_NDVI,
         SI = uas NDVI SI,
         RI = uas_NDVI_Response_Index)
paper3_uas_ndre_data <- paper3_data %>%
  dplyr::select(site_year ,
                year ,
                exp_plot_number ,
                Block ,
                MainPlot ,
                SubPlot ,
                N_level_kgha ,
                N_level_kgha_f ,
                TopDress_kgha ,
                TopDress_kgha_f ,
                uas_NDRE ,
                uas_NDRE_Response_Index ,
                uas_NDRE_SI ,
                GrainYield Mgha,
                PI_N_Uptake) %>%
  mutate(Platform = "sUAS NDRE" ) %>%
  rename(Index = uas_NDRE ,
         SI = uas_NDRE_SI,
         RI = uas_NDRE_Response_Index)
paper3_data <- rbind(paper3_gsdata ,</pre>
                     paper3_uas_ndvi_data ,
                     paper3_uas_ndre_data)
paper3_data <- paper3_data %>%
 dplyr::select(site_year ,
                year ,
                Platform,
                exp_plot_number ,
                Block ,
                MainPlot,
                SubPlot ,
                N_level_kgha ,
                N_level_kgha_f ,
```

```
TopDress_kgha
               TopDress_kgha_f ,
               Index,
               RI,
               SI.
               GrainYield_Mgha,
               PI_N_Uptake)
paper3_data$Platform <- as.factor(paper3_data$Platform)</pre>
str(paper3_data , give.attr = F)
## rowwise_df [720 x 16] (S3: rowwise_df/tbl_df/tbl/data.frame)
## $ site_year : Factor w/ 10 levels "Nicolaus-17",..: 1 1 1 1 1 1 1 1 1 1 ...
                    : Factor w/ 3 levels "2017", "2018", ...: 1 1 1 1 1 1 1 1 1 1 ...
## $ year
                 : Factor w/ 3 levels "GreenSeeker NDVI",..: 1 1 1 1 1 1 1 1 1 1 ...
## $ Platform
## $ exp_plot_number: Factor w/ 27 levels "101","102","103",..: 1 2 3 4 6 7 9 10 11 12 ...
                   : Factor w/ 4 levels "1","2","3","4": 1 1 1 1 1 2 2 2 2 ...
## $ Block
## $ MainPlot
                   : Factor w/ 7 levels "1","2","3","4",..: 1 2 3 4 6 7 2 3 4 5 ...
                : Factor w/ 3 levels "1","2","3": 2 3 1 1 1 2 1 1 2 3 ...
## $ SubPlot
## $ N level kgha : num [1:720] 225 0 125 175 45 75 75 125 0 175 ...
## $ N level kgha f : Factor w/ 11 levels "0","45","75",...: 10 1 5 8 2 3 3 5 1 8 ...
## $ TopDress_kgha : num [1:720] 0 0 0 0 0 0 0 0 0 ...
## $ TopDress kgha f: Factor w/ 1 level "0": 1 1 1 1 1 1 1 1 1 1 1 ...
## $ Index
                   : num [1:720] 0.777 0.493 0.683 0.787 0.683 ...
## $ RI
                    : num [1:720] 1.012 1.593 1.15 0.999 1.15 ...
## $ SI
                    : num [1:720] 0.988 0.628 0.87 1.001 0.87 ...
## $ GrainYield_Mgha: num [1:720] 11.4 11 12 11.9 11.5 ...
## $ PI_N_Uptake
                   : num [1:720] 158 78 142 197 95 ...
paper3_data <- tibble::rowid_to_column(paper3_data, "ID") #adds a columns with row number.</pre>
```

Calculating Relative Yield

```
nic17 <- subset(paper3_data, site_year == "Nicolaus-17")
nic17 <- apply(nic17,2,max)
nic17max_yield <- as.numeric(nic17[16])

wil17 <- subset(paper3_data, site_year == "Williams-17")
wil17 <- apply(wil17,2,max)
wil17max_yield <- as.numeric(wil17[16])

arb18 <- subset(paper3_data, site_year == "Arbuckle-18")
arb18 <- apply(arb18,2,max)
arb18max_yield <- as.numeric(arb18[16])

nic18 <- subset(paper3_data, site_year == "Nicolaus-18")
nic18 <- apply(nic18,2,max)
nic18max_yield <- as.numeric(nic18[16])

mry18 <- subset(paper3_data, site_year == "Marysville-18")
mry18 <- apply(mry18,2,max)
mry18max_yield <- as.numeric(mry18[16])</pre>
```

```
biggs18 <- subset(paper3_data, site_year == "Biggs-18")</pre>
biggs18 <- apply(biggs18,2,max)</pre>
biggs18max_yield <- as.numeric(biggs18[16])</pre>
arb19 <- subset(paper3_data, site_year == "Arbuckle-19")</pre>
arb19 <- apply(arb19,2,max)</pre>
arb19max_yield <- as.numeric(arb19[16])</pre>
dav19 <- subset(paper3_data, site_year == "Davis-19")</pre>
dav19 <- apply(dav19,2,max)</pre>
dav19max_yield <- as.numeric(dav19[16])</pre>
mry19 <- subset(paper3_data, site_year == "Marysville-19")</pre>
mry19 <- apply(mry19,2,max)</pre>
mry19max_yield <- as.numeric(mry19[16])</pre>
res19 <- subset(paper3_data, site_year == "RES-19")
res19 <- apply(res19,2,max)</pre>
res19max_yield <- as.numeric(res19[16])</pre>
paper3_data <- paper3_data %>%
  mutate(relative_grain_yield = case_when(
                            site_year == "Nicolaus-17" ~ GrainYield_Mgha / nic17max_yield ,
                            site_year == "Williams-17" ~ GrainYield_Mgha / wil17max_yield ,
                            site_year == "Arbuckle-18" ~ GrainYield_Mgha / arb18max_yield ,
                            site_year == "Nicolaus-18" ~ GrainYield_Mgha / nic18max_yield ,
                            site_year == "Marysville-18" ~ GrainYield_Mgha / mry18max_yield ,
                           site_year == "Biggs-18" ~ GrainYield_Mgha / biggs18max_yield,
                           site_year == "Arbuckle-19" ~ GrainYield_Mgha / arb19max_yield ,
                            site_year == "Davis-19" ~ GrainYield_Mgha / dav19max_yield ,
                            site_year == "Marysville-19" ~ GrainYield_Mgha / mry19max_yield ,
                            site_year == "RES-19" ~ GrainYield_Mgha / res19max_yield))
```

Outlier Removal

```
filter1 <- paper3_data %>%
    filter(site_year == "Biggs-18" & exp_plot_number == 101)

paper3_data <- paper3_data %>%
    filter(!ID %in% filter1$ID) #removes Biggs-18 plot 101 plot bc tractor ran through it

paper3_data$Platform = factor(paper3_data$Platform, levels=c( "GreenSeeker NDVI" , "sUAS NDRE" , "sUAS

hist_data <- paper3_data %>%
    dplyr::select(Platform , SI) %>%
    group_by(Platform) %>%
    summarise(mean = mean(SI) , sd = sd(SI) , median = median(SI))

hist_data_gs <- hist_data %>%
    filter(Platform == "GreenSeeker NDVI")
```

```
lower_limit <- hist_data_gs[2] - (4*hist_data_gs[3]) #upper limit for outlier removal. Observations tha
print(lower_limit)
##
          mean
## 1 0.1997943
paper3_gsdata <- paper3_data %>%
  filter(Platform == "GreenSeeker NDVI")
paper3_uas_ndvi_data <- paper3_data %>%
  filter(Platform == "sUAS NDVI")
paper3_uas_ndre_data <- paper3_data %>%
 filter(Platform == "sUAS NDRE")
round(tapply(paper3_gsdata$GrainYield_Mgha , paper3_gsdata$site_year , max) , digits = 1)
##
     Nicolaus-17 Williams-17
                                 Arbuckle-18
                                                  Biggs-18 Marysville-18
##
            12.9
                          12.1
                                        13.1
                                                       12.6
                                                                     12.0
                                                                  RES-19
    Nicolaus-18
                  Arbuckle-19
                                    Davis-19 Marysville-19
##
##
            13.8
                          12.0
                                        11.7
                                                       11.3
                                                                      9.6
min(paper3_uas_ndvi_data$SI)
## [1] 0.6349517
max(paper3_uas_ndvi_data$SI)
## [1] 1.041521
min(paper3 gsdata$SI)
## [1] 0.1985336
max(paper3_gsdata$SI)
## [1] 1.133537
min(paper3_uas_ndre_data$SI)
## [1] 0.4903767
max(paper3_uas_ndre_data$SI)
## [1] 1.101968
round((length(which(paper3_uas_ndvi_data$SI >= .85)) / length(paper3_uas_ndvi_data$SI)) * 100 , digits
## [1] 92
round((length(which(paper3_uas_ndre_data$SI >= .85)) / length(paper3_uas_ndre_data$SI)) * 100 , digits
## [1] 74
round((length(which(paper3_gsdata$SI >= .85)) / length(paper3_gsdata$SI)) * 100 , digits = 0)
## [1] 73
test <- paper3_uas_ndre_data %>%
  filter(SI >= .90)
```

```
test <- round((sd(test$relative_grain_yield)*100) , digits = 1)</pre>
test
## [1] 5.9
test <- paper3_uas_ndvi_data %>%
  filter(SI >= .90)
test <- round((sd(test$relative_grain_yield)*100) , digits = 1)</pre>
test
## [1] 9.6
test <- paper3_gsdata %>%
  filter(SI >= .90)
test <- round((sd(test$relative_grain_yield)*100) , digits = 1)</pre>
test
## [1] 6.5
round((length(which(paper3_uas_ndvi_data$SI >= .9)) / length(paper3_uas_ndvi_data$SI)) * 100 , digits =
## [1] 87.4
round((length(which(paper3_uas_ndre_data$SI >= .9)) / length(paper3_uas_ndre_data$SI)) * 100 , digits =
round((length(which(paper3_gsdata$SI >= .9)) / length(paper3_gsdata$SI)) * 100 , digits = 1)
## [1] 65.7
FIGURE 2
Data
suppl_data <- paper3_data</pre>
suppl_data1 <- suppl_data %>%
  {\tt dplyr::select(site\_year, N\_level\_kgha \ , \ TopDress\_kgha\_f \ , \ GrainYield\_Mgha \ , \ PI\_N\_Uptake) \ \%>\%}
  group_by(site_year , N_level_kgha ,TopDress_kgha_f ) %>%
  summarise_all(.funs = c(mean , sd)) %>%
  rename(PI_N_Uptake = PI_N_Uptake_fn1,
         GrainYield_Mgha = GrainYield_Mgha_fn1) %>%
  mutate(PI_N_Uptake_r = round(PI_N_Uptake , digits = 0),
         GrainYield_Mgha_r = round(GrainYield_Mgha , digits = 1)) %>%
  ungroup()
```

site
Nicolaus-17 Williams-17 Arbuckle-18 Biggs-18 Marysville-18

factor_list <- list(site = suppl_data1\$site_year)</pre>

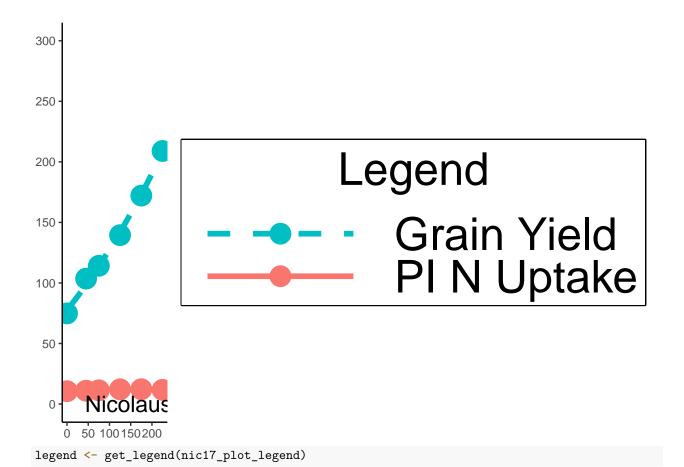
tapply(suppl_data1\$GrainYield_Mgha_r, factor_list , max)

```
12.2
                                          12.7
##
                           11.1
                                                         12.3
                                                                        11.3
                   Arbuckle-19
##
     Nicolaus-18
                                     Davis-19 Marysville-19
                                                                     RES-19
                           11.5
                                          10.7
##
            13.3
                                                                        9.1
tapply(suppl_data1$GrainYield_Mgha_r, factor_list , min)
## site
##
     Nicolaus-17
                   Williams-17
                                   Arbuckle-18
                                                    Biggs-18 Marysville-18
##
                            6.6
                                           3.1
                                                          7.3
                                                                        8.7
            10.6
     Nicolaus-18
                                     Davis-19 Marysville-19
                                                                     RES-19
##
                   Arbuckle-19
             9.2
                            4.3
                                           7.6
                                                          6.9
                                                                        6.0
##
suppl data2 <- suppl data %>%
  dplyr::select(site_year, N_level_kgha , TopDress_kgha_f , Platform , Index) %>%
  group_by(site_year , N_level_kgha ,TopDress_kgha_f , Platform ) %>%
  summarise_all(.funs = c(mean , sd)) %>%
  rename(Index = fn1,
         sd = fn2) \% \%
  ungroup()
ggplot(data = suppl_data2 , aes ( x = N_level_kgha , Index , color = Platform)) +
  geom_point(data = suppl_data2 , aes ( x = N_level_kgha , Index , color = Platform)) +
  coord_cartesian(ylim = c(0.2,1)) +
  geom_smooth(method = "lm", formula = y ~ poly(x, 2), se = FALSE) +
  theme_classic() +
  facet_wrap(~site_year)
        Nicolaus-17
                        Williams-17
                                        Arbuckle-18
                                                        Biggs-18
  1.0
  0.8
  0.6
  0.4
  0.2
       Marysville-18
                        Nicolaus-18
                                       Arbuckle-19
                                                        Davis-19
  1.0
                                                                       Platform
  0.8
                                                                       GreenSeeker NDVI
  0.6
                                                                           sUAS NDRE
  0.4
                                                                           sUAS NDVI
  0.2
                                       50 100150200 0 50 100150200
        Marysville-19
                         RES-19
  1.0
  0.8
  0.6
  0.4
  0.2
      0 50 100150200 0 50 100150200
```

N_level_kgha

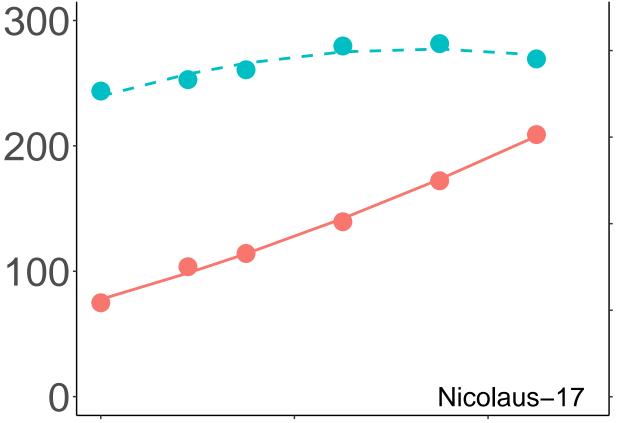
Legend

```
suppl_data1_nic17 <- suppl_data1 %>%
 filter(site year == "Nicolaus-17")
suppl_data1_nic17_yield <- suppl_data1_nic17 %>%
  dplyr::select(-c(PI_N_Uptake)) %>%
  mutate(Legend = as.factor("Grain Yield")) %>%
  rename("Unit" = GrainYield_Mgha)
suppl_data1_nic17_nup <- suppl_data1_nic17 %>%
  dplyr::select(-c(GrainYield_Mgha)) %>%
  mutate(Legend = as.factor("PI N Uptake")) %>%
  rename("Unit" = PI_N_Uptake)
suppl_data1_nic17 <- rbind(suppl_data1_nic17_nup , suppl_data1_nic17_yield)</pre>
nic17_plot_legend <- ggplot(data = suppl_data1_nic17 , aes( x = N_level_kgha , y = Unit , color = Legen
  geom_point(data = suppl_data1_nic17, aes (x = N_level_kgha, y = Unit, color = Legend), size = 7)
  scale_y_continuous(breaks = seq(0, 300, by = 50)) +
  coord_cartesian(ylim = c(0 , 300)) +
  theme_classic() +
  theme(legend.text = element_text(size = 34 ),
        legend.title = element_text(size = 34 ),
        legend.box.background = element_rect(size = 1) ) +
  annotate("text", x = 190, y = 0, label = "Nicolaus-17", size = 6) +
  labs(x = NULL, y = NULL) +
  geom\_smooth(method = "lm", formula = y ~ poly(x, 2), se = FALSE, size = 2) +
  guides(color = guide_legend(keywidth = 9.5 , keyheight = 1.5 , unit = "cm" , title.hjust = 0.5)) +
  scale_color_manual("Legend" , labels = c("Grain Yield" , "PI N Uptake") , values = c( "#00BFC4" , "#F
  scale_linetype_manual("Legend" , labels = c("Grain Yield" , "PI N Uptake") , values = c( "dashed" , "
nic17_plot_legend
```



Nicolaus-17

```
suppl_data1_nic17 <- suppl_data1 %>%
  filter(site_year == "Nicolaus-17")
nic17_nuptake_lm <- lm(PI_N_Uptake ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_data1_n
nic17_fitted <- fitted(nic17_nuptake_lm)</pre>
nic17_df <- data.frame(suppl_data1_nic17$N_level_kgha , nic17_fitted) #creates dataframe
nic17_yield_lm <- lm(GrainYield_Mgha ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_data1
nic17_a <- as.numeric(as.character(coef(nic17_yield_lm)[3]))</pre>
nic17_b <- as.numeric(as.character(coef(nic17_yield_lm)[2]))</pre>
nic17_c <- as.numeric(as.character(coef(nic17_yield_lm)[1]))</pre>
mic17sym_x \leftarrow round((-mic17_b) / (2*mic17_a) , digits = 0)
nic17sym_y <- round(nic17_a*(nic17sym_x^2) + nic17_b*nic17sym_x + nic17_c , digits = 1)</pre>
nic17_fitted_2 <- (fitted(nic17_yield_lm)*23)</pre>
nic17_df_2 <- data.frame(suppl_data1_nic17$N_level_kgha , nic17_fitted_2) #creates dataframe
mic17_plot \leftarrow ggplot(data = suppl_data1_mic17, aes(x = N_level_kgha, y = PI_N_Uptake)) +
  geom_point(data = suppl_data1_nic17, aes (x = N_level_kgha, y = PI_N_Uptake), color = "#F8766D",
  geom_point(\frac{data}{ata} = suppl_data1_nic17, aes (x = N_level_kgha, y = GrainYield_Mgha*23), color = "#000]
  geom_line(data = nic17_df , aes( x = suppl_data1_nic17.N_level_kgha , y = nic17_fitted) , color = "#F
```



Williams-17

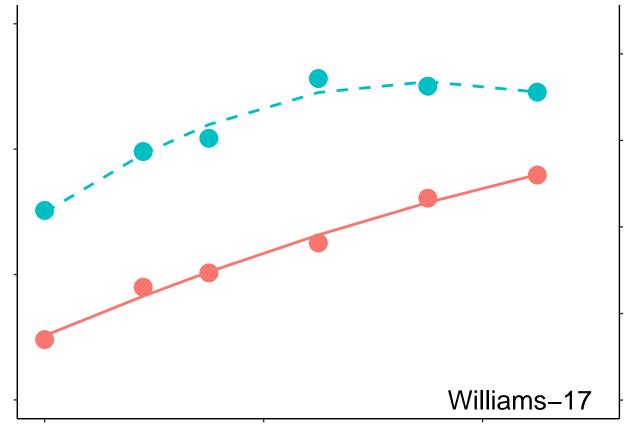
```
suppl_data1_wil17 <- suppl_data1 %>%
  filter(site_year == "Williams-17")

wil17_nuptake_lm <- lm(PI_N_Uptake ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_data1_w

wil17_fitted <- fitted(wil17_nuptake_lm)
wil17_df <- data.frame(suppl_data1_wil17$N_level_kgha , wil17_fitted) #creates dataframe

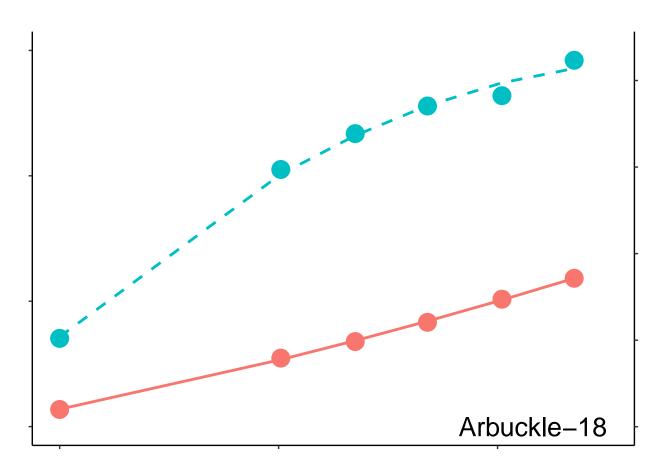
wil17_yield_lm <- lm(GrainYield_Mgha ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_data1</pre>
```

```
wil17_a <- as.numeric(as.character(coef(wil17_yield_lm)[3]))</pre>
wil17_b <- as.numeric(as.character(coef(wil17_yield_lm)[2]))</pre>
wil17_c <- as.numeric(as.character(coef(wil17_yield_lm)[1]))</pre>
will7sym_x \leftarrow round((-will7_b) / (2*will7_a), digits = 0)
 will 17 sym_y \leftarrow round(will 17_a*(will 17 sym_x^2) + will 17_b*will 17 sym_x + will 17_c , \frac{digits}{} = 1) 
wil17_fitted_2 <- (fitted(wil17_yield_lm)*23)</pre>
wil17_df_2 <- data.frame(suppl_data1_wil17$N_level_kgha , wil17_fitted_2) #creates dataframe
wil17_plot \leftarrow ggplot(data = suppl_data1_wil17, aes(x = N_level_kgha, y = PI_N_Uptake)) +
     geom_point(data = suppl_data1_wil17 , aes ( x = N_level_kgha , y = PI_N_Uptake) , color = "#F8766D" ,
     geom_point(data = suppl_data1_wil17 , aes ( x = N_level_kgha , y = GrainYield_Mgha*23) , color = "#00]
     geom_line(data = wil17_df , aes( x = suppl_data1_wil17.N_level_kgha , y = wil17_fitted) , color = "#F
     geom\_line(\frac{data}{data} = wil17\_df\_2 , aes( x = suppl\_data1\_wil17.N\_level\_kgha , y = wil17\_fitted\_2) , color = wil17\_fit
     scale_y = continuous(breaks = seq(0, 300, by = 100), sec.axis = sec_axis(~./23, breaks = seq(0, 1))
     coord_cartesian(ylim = c(0, 300), xlim = c(0, 250)) +
     scale_x_continuous(breaks = seq(0 , 280 , by = 100)) +
     theme_classic() +
     theme(axis.text.y = element_blank(),
                    axis.text.x = element_blank()) +
     annotate("text", x = 250, y = 0, label = "Williams-17", size = 7, hjust = 1) +
     labs(x = NULL, y = NULL)
wil17_plot
```



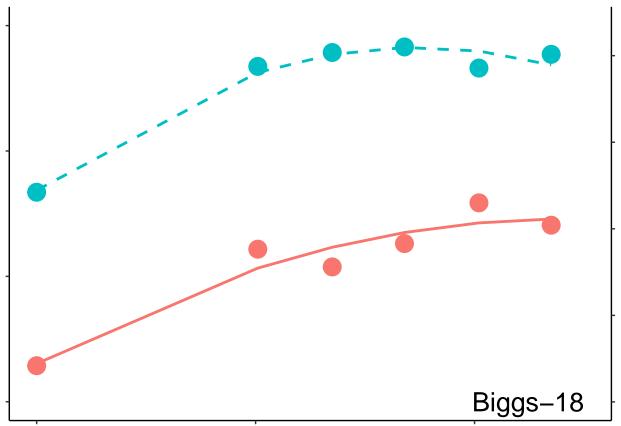
Arbuckle-18

```
suppl_data1_arb18 <- suppl_data1 %>%
 filter(site year == "Arbuckle-18")
arb18_nuptake_lm <- lm(PI_N_Uptake ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_data1_a
arb18_fitted <- fitted(arb18_nuptake_lm)</pre>
arb18_df <- data.frame(suppl_data1_arb18$N_level_kgha , arb18_fitted) #creates dataframe
arb18_yield_lm <- lm(GrainYield_Mgha ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_data1
arb18_a <- as.numeric(as.character(coef(arb18_yield_lm)[3]))</pre>
arb18_b <- as.numeric(as.character(coef(arb18_yield_lm)[2]))</pre>
arb18_c <- as.numeric(as.character(coef(arb18_yield_lm)[1]))</pre>
arb18sym_x \leftarrow round((-arb18_b) / (2*arb18_a) , digits = 0)
arb18sym_y <- round(arb18_a*(arb18sym_x^2) + arb18_b*arb18sym_x + arb18_c , digits = 1)
arb18_fitted_2 <- (fitted(arb18_yield_lm)*23)</pre>
arb18_df_2 <- data.frame(suppl_data1_arb18$N_level_kgha , arb18_fitted_2) #creates dataframe
arb18_plot \leftarrow ggplot(data = suppl_data1_arb18 , aes(x = N_level_kgha , y = PI_N_Uptake)) +
  geom_point(data = suppl_data1_arb18 , aes ( x = N_level_kgha , y = PI_N_Uptake) , color = "#F8766D" ,
  geom_point(data = suppl_data1_arb18 , aes (x = N_level_kgha , y = GrainYield_Mgha*23) , color = "#00"
  geom_line(data = arb18_df , aes( x = suppl_data1_arb18.N_level_kgha , y = arb18_fitted) , color = "#F
  geom_line(data = arb18_df_2 , aes( x = suppl_data1_arb18.N_level_kgha , y = arb18_fitted_2) , color =
  scale_y = seq(0, 300, by = 100), sec.axis = sec_axis(~./23, breaks = seq(0, 1)
  coord_cartesian(ylim = c(0 , 300) , xlim = c(0 , 250)) +
  scale_x_continuous(breaks = seq(0, 280, by = 100)) +
  theme_classic() +
  theme(axis.text.y = element_blank(),
        axis.text.x = element_blank()) +
  annotate("text" , x = 250 , y = 0 , label = "Arbuckle-18" , size = 7 , hjust = 1) +
  labs(x = NULL, y = NULL)
arb18_plot
```



Biggs-18

```
suppl_data1_biggs18 <- suppl_data1 %>%
  filter(site_year == "Biggs-18" )
biggs18_nuptake_lm <- lm(PI_N_Uptake ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_data1
biggs18_fitted <- fitted(biggs18_nuptake_lm)</pre>
biggs18_df <- data.frame(suppl_data1_biggs18$N_level_kgha , biggs18_fitted) #creates dataframe
biggs18_yield_lm <- lm(GrainYield_Mgha ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_dat
biggs18_a <- as.numeric(as.character(coef(biggs18_yield_lm)[3]))</pre>
biggs18_b <- as.numeric(as.character(coef(biggs18_yield_lm)[2]))</pre>
biggs18_c <- as.numeric(as.character(coef(biggs18_yield_lm)[1]))</pre>
biggs18sym_x \leftarrow round((-biggs18_b) / (2*biggs18_a) , digits = 0)
biggs18sym_y <- round(biggs18_a*(biggs18sym_x^2) + biggs18_b*biggs18sym_x + biggs18_c , digits = 1)
biggs18_fitted_2 <- (fitted(biggs18_yield_lm)*23)</pre>
biggs18_df_2 <- data.frame(suppl_data1_biggs18$N_level_kgha , biggs18_fitted_2) #creates dataframe
biggs18\_plot \leftarrow ggplot(\frac{data}{ata} = suppl\_data1\_biggs18 , aes( x = N\_level\_kgha , y = PI\_N\_Uptake)) +
  geom\_point(data = suppl\_data1\_biggs18 , aes ( x = N\_level\_kgha , y = PI\_N\_Uptake) , color = "\#F8766D"
  geom_point(data = suppl_data1_biggs18 , aes ( x = N_level_kgha , y = GrainYield_Mgha*23) , color = "#"
  geom\_line(data = biggs18\_df , aes(x = suppl\_data1\_biggs18.N\_level\_kgha , y = biggs18\_fitted) , color
  geom_line(data = biggs18_df_2 , aes( x = suppl_data1_biggs18.N_level_kgha , y = biggs18_fitted_2) , c
```



Marysville-18

```
suppl_data1_mry18 <- suppl_data1 %>%
    filter(site_year == "Marysville-18" )

mry18_nuptake_lm <- lm(PI_N_Uptake ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_data1_m

mry18_fitted <- fitted(mry18_nuptake_lm)

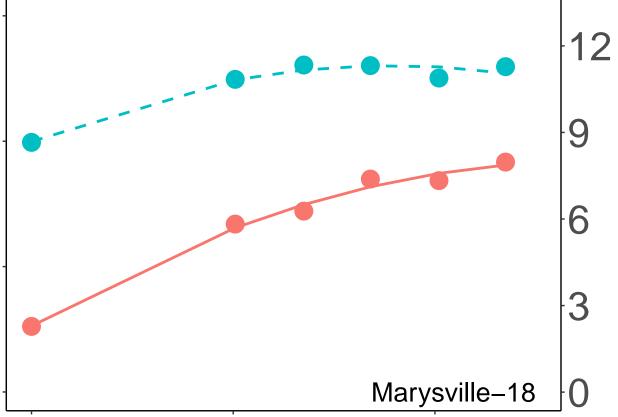
mry18_df <- data.frame(suppl_data1_mry18$N_level_kgha , mry18_fitted) #creates dataframe

mry18_yield_lm <- lm(GrainYield_Mgha ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_data1

mry18_a <- as.numeric(as.character(coef(mry18_yield_lm)[3]))

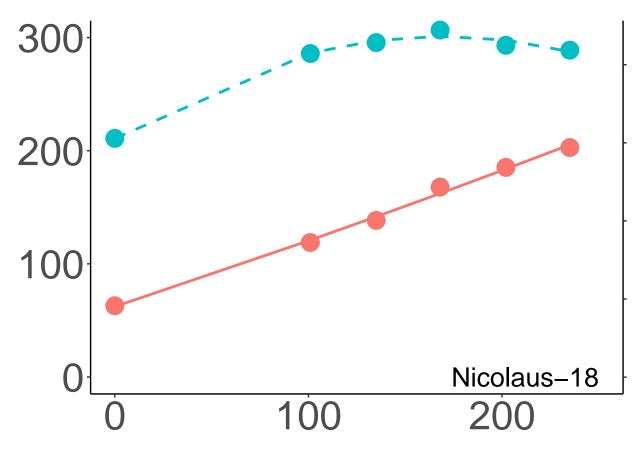
mry18_b <- as.numeric(as.character(coef(mry18_yield_lm)[2]))</pre>
```

```
mry18_c <- as.numeric(as.character(coef(mry18_yield_lm)[1]))</pre>
mry18sym_x <- round((-mry18_b) / (2*mry18_a) , digits = 0)</pre>
mry18sym_y \leftarrow round(mry18_a*(mry18sym_x^2) + mry18_b*mry18sym_x + mry18_c , digits = 1)
mry18_fitted_2 <- (fitted(mry18_yield_lm)*23)</pre>
mry18_df_2 <- data.frame(suppl_data1_mry18$N_level_kgha , mry18_fitted_2) #creates dataframe
mry18_plot \leftarrow ggplot(data = suppl_data1_mry18 , aes(x = N_level_kgha , y = PI_N_Uptake)) +
        geom_point(data = suppl_data1_mry18 , aes (x = N_level_kgha , y = PI_N_Uptake) , color = "#F8766D" ,
        geom_point(\frac{data}{a} = suppl_data1_mry18 , aes ( x = N_level_kgha , y = GrainYield_Mgha*23) , color = "#000" | Supplementation | Supplementaries | Supple
        geom_line(data = mry18_df , aes( x = suppl_data1_mry18.N_level_kgha , y = mry18_fitted) , color = "#F
        geom\_line(data = mry18\_df\_2 , aes(x = suppl\_data1\_mry18.N\_level\_kgha , y = mry18\_fitted\_2) , color = mry18\_fitted\_2) , c
        scale_y_continuous(breaks = seq(0, 300, by = 100), sec.axis = sec_axis(~./23, breaks = seq(0, 100))
        coord_cartesian(ylim = c(0 , 300) , xlim = c(0 , 250)) +
        scale_x_continuous(breaks = seq(0, 280, by = 100)) +
        theme_classic() +
        theme(axis.text.y.right = element_text(size = 30),
                               axis.text.y.left = element_blank(),
                                axis.text.x = element_blank()) +
        annotate("text", x = 250, y = 0, label = "Marysville-18", size = 7, hjust = 1) +
        labs(x = NULL, y = NULL)
mry18_plot
```



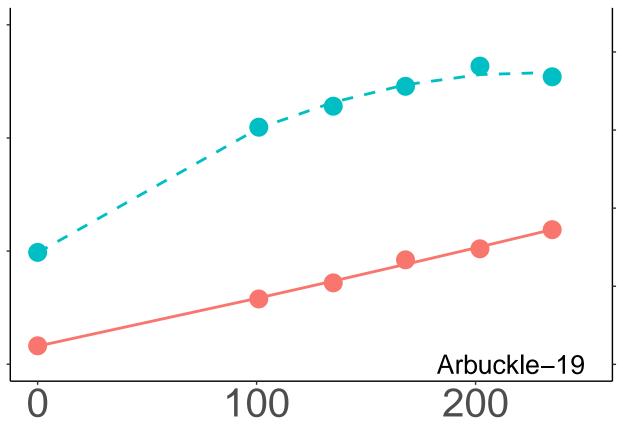
Nicolaus-18

```
suppl_data1_nic18 <- suppl_data1 %>%
  filter(site year == "Nicolaus-18" )
nic18_nuptake_lm <- lm(PI_N_Uptake ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_data1_n
nic18_fitted <- fitted(nic18_nuptake_lm)</pre>
nic18_df <- data.frame(suppl_data1_nic18$N_level_kgha , nic18_fitted) #creates dataframe
nic18_yield_lm <- lm(GrainYield_Mgha ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_data1
nic18_a <- as.numeric(as.character(coef(nic18_yield_lm)[3]))</pre>
nic18_b <- as.numeric(as.character(coef(nic18_yield_lm)[2]))</pre>
nic18_c <- as.numeric(as.character(coef(nic18_yield_lm)[1]))</pre>
nic18sym_x \leftarrow round((-nic18_b) / (2*nic18_a) , digits = 0)
nic18sym_y <- round(nic18_a*(nic18sym_x^2) + nic18_b*nic18sym_x + nic18_c , digits = 1)
nic18_fitted_2 <- (fitted(nic18_yield_lm)*23)</pre>
nic18_df_2 <- data.frame(suppl_data1_nic18$N_level_kgha , nic18_fitted_2) #creates dataframe
mic18_plot \leftarrow ggplot(\frac{data}{a} = suppl_data1_mic18, aes(x = N_level_kgha, y = PI_N_Uptake)) +
  geom_point(data = suppl_data1_nic18 , aes ( x = N_level_kgha , y = PI_N_Uptake) , color = "#F8766D" ,
  geom_point(data = suppl_data1_nic18 , aes (x = N_level_kgha , y = GrainYield_Mgha*23) , color = "#00"
  geom_line(data = nic18_df , aes( x = suppl_data1_nic18.N_level_kgha , y = nic18_fitted) , color = "#F
  geom_line(data = nic18_df_2 , aes( x = suppl_data1_nic18.N_level_kgha , y = nic18_fitted_2) , color =
  scale_y = seq(0, 300, by = 100), sec.axis = sec_axis(~./23, breaks = seq(0, 1)
  coord_cartesian(ylim = c(0, 300), xlim = c(0, 250)) +
  scale_x_continuous(breaks = seq(0, 280, by = 100)) +
  theme_classic() +
  theme(axis.text.y.right = element_blank(),
        axis.text.x = element_text(size = 30),
        axis.text.y.left = element_text(size = 30)) +
  annotate("text", x = 250, y = 0, label = "Nicolaus-18", size = 7, hjust = 1) +
  labs(x = NULL, y = NULL)
nic18 plot
```



Arbuckle-19

```
suppl_data1_arb19 <- suppl_data1 %>%
    filter(site_year == "Arbuckle-19" )
arb19_nuptake_lm <- lm(PI_N_Uptake ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_data1_a
arb19_fitted <- fitted(arb19_nuptake_lm)</pre>
arb19 df <- data.frame(suppl data1 arb19$N level kgha , arb19 fitted) #creates dataframe
arb19_yield_lm <- lm(GrainYield_Mgha ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_data1
arb19_a <- as.numeric(as.character(coef(arb19_yield_lm)[3]))</pre>
arb19_b <- as.numeric(as.character(coef(arb19_yield_lm)[2]))</pre>
arb19_c <- as.numeric(as.character(coef(arb19_yield_lm)[1]))</pre>
arb19sym_x \leftarrow round((-arb19_b) / (2*arb19_a) , digits = 0)
arb19sym_y \leftarrow round(arb19_a*(arb19sym_x^2) + arb19_b*arb19sym_x + arb19_c , digits = 1)
arb19_fitted_2 <- (fitted(arb19_yield_lm)*23)</pre>
arb19_df_2 <- data.frame(suppl_data1_arb19$N_level_kgha , arb19_fitted_2) #creates dataframe
arb19_plot \leftarrow ggplot(data = suppl_data1_arb19 , aes( x = N_level_kgha , y = PI_N_Uptake)) +
    geom\_point(data = suppl\_data1\_arb19 , aes ( x = N\_level\_kgha , y = PI\_N\_Uptake) , color = "#F8766D" ,
    geom\_point(data = suppl\_data1\_arb19 , aes ( x = N\_level\_kgha , y = GrainYield\_Mgha*23) , color = "#000 | for the color | for
    geom_line(data = arb19_df , aes( x = suppl_data1_arb19.N_level_kgha , y = arb19_fitted) , color = "#F
    geom_line(data = arb19_df_2 , aes( x = suppl_data1_arb19.N_level_kgha , y = arb19_fitted_2) , color =
    scale_y\_continuous(breaks = seq(0, 300, by = 100), sec.axis = sec_axis(~./23, breaks = seq(0, 100))
```



Davis-19

```
suppl_data1_davis19 <- suppl_data1 %>%
    filter(site_year == "Davis-19" )

davis19_nuptake_lm <- lm(PI_N_Uptake ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_data1
davis19_fitted <- fitted(davis19_nuptake_lm)
davis19_df <- data.frame(suppl_data1_davis19$N_level_kgha , davis19_fitted) #creates dataframe

davis19_yield_lm <- lm(GrainYield_Mgha ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_dat
davis19_a <- as.numeric(as.character(coef(davis19_yield_lm)[3]))
davis19_b <- as.numeric(as.character(coef(davis19_yield_lm)[2]))
davis19_c <- as.numeric(as.character(coef(davis19_yield_lm)[1]))
davis19sym_x <- (-davis19_b) / (2*davis19_a)</pre>
```

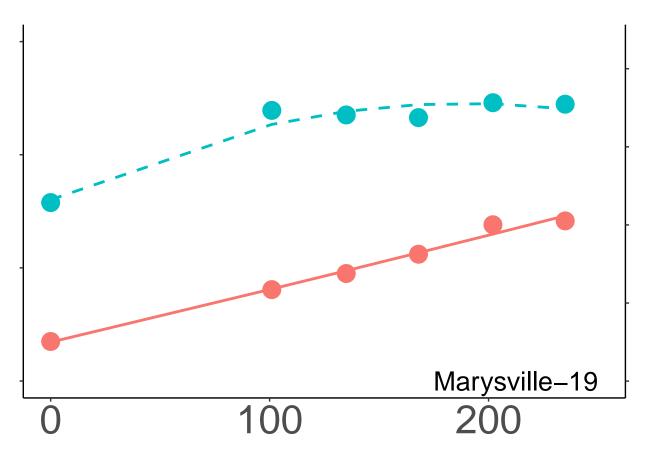
```
davis19sym_y <- davis19_a*(davis19sym_x^2) + davis19_b*davis19sym_x + davis19_c
davis19_fitted_2 <- (fitted(davis19_yield_lm)*23)</pre>
davis19_df_2 <- data.frame(suppl_data1_davis19$N_level_kgha , davis19_fitted_2) #creates dataframe
geom_point(data = suppl_data1_davis19), aes ( x = N_level_kgha, y = PI_N_Uptake), color = "#F8766D"
 geom_point(data = suppl_data1_davis19), aes ( x = N_level_kgha, y = GrainYield_Mgha*23), color = "#
 geom\_line(data = davis19\_df , aes(x = suppl\_data1\_davis19.N\_level\_kgha , y = davis19\_fitted) , color
 geom\_line(data = davis19\_df\_2 , aes(x = suppl\_data1\_davis19.N\_level\_kgha , y = davis19\_fitted\_2) , c
 scale_y = continuous(breaks = seq(0, 300, by = 100), sec.axis = sec_axis(~./23, breaks = seq(0, 1))
 coord_cartesian(ylim = c(0 , 300) , xlim = c(0 , 250)) +
 scale x continuous(breaks = seq(0, 280, by = 100)) +
 theme_classic() +
 theme(axis.text.y = element_blank(),
       axis.text.x = element_text(size = 30))+
 annotate("text" , x = 250 , y = 0 , label = "Davis-19" , size = 7 , hjust = 1) +
 labs(x = NULL, y = NULL)
davis19_plot
```

Marysville-19

```
suppl_data1_mry19 <- suppl_data1 %>%
filter(site_year == "Marysville-19" )
```

Davis-19

```
mry19_nuptake_lm <- lm(PI_N_Uptake ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_data1_m
mry19_fitted <- fitted(mry19_nuptake_lm)</pre>
mry19_df <- data.frame(suppl_data1_mry19$N_level_kgha , mry19_fitted) #creates dataframe
mry19_yield_lm <- lm(GrainYield_Mgha ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_data1
mry19 a <- as.numeric(as.character(coef(mry19 yield lm)[3]))</pre>
mry19_b <- as.numeric(as.character(coef(mry19_yield_lm)[2]))</pre>
mry19_c <- as.numeric(as.character(coef(mry19_yield_lm)[1]))</pre>
mry19sym_x <- round((-mry19_b) / (2*mry19_a) , digits = 0)</pre>
mry19sym_y <- round(mry19_a*(mry19sym_x^2) + mry19_b*mry19sym_x + mry19_c , digits = 1)</pre>
mry19_fitted_2 <- (fitted(mry19_yield_lm)*23)</pre>
mry19_df_2 <- data.frame(suppl_data1_mry19$N_level_kgha , mry19_fitted_2) #creates dataframe
mry19_plot \leftarrow ggplot(data = suppl_data1_mry19 , aes(x = N_level_kgha , y = PI_N_Uptake)) +
    geom_point(data = suppl_data1_mry19 , aes ( x = N_level_kgha , y = PI_N_Uptake) , color = "#F8766D" ,
    geom_point(data = suppl_data1_mry19), aes ( x = N_level_kgha , y = GrainYield_Mgha*23) , color = "#00"
    geom_line(data = mry19_df , aes( x = suppl_data1_mry19.N_level_kgha , y = mry19_fitted) , color = "#F
    geom\_line(data = mry19\_df\_2 , aes(x = suppl\_data1\_mry19.N\_level\_kgha , y = mry19\_fitted\_2) , color = mry10\_fitted\_2) , c
    scale_y = continuous(breaks = seq(0, 300, by = 100), sec.axis = sec_axis(~./23, breaks = seq(0, 1))
    coord_cartesian(ylim = c(0, 300), xlim = c(0, 250)) +
         scale_x_continuous(breaks = seq(0, 280, by = 100)) +
    theme classic() +
    theme(axis.text.x = element_text(size = 30),
                 axis.text.y = element blank()) +
    annotate("text" , x = 250 , y = 0 , label = "Marysville-19" , size = 7 , hjust = 1) +
    labs(x = NULL, y = NULL)
mry19_plot
```



RES-19

```
suppl_data1_res19 <- suppl_data1 %>%
     filter(site_year == "RES-19" )
res19_nuptake_lm <- lm(PI_N_Uptake ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_data1_r
res19_fitted <- fitted(res19_nuptake_lm)</pre>
res19_df <- data.frame(suppl_data1_res19$N_level_kgha , res19_fitted) #creates dataframe
res19_yield_lm <- lm(GrainYield_Mgha ~ N_level_kgha + I(N_level_kgha*N_level_kgha) , data = suppl_data1
res19_a <- as.numeric(as.character(coef(res19_yield_lm)[3]))</pre>
res19_b <- as.numeric(as.character(coef(res19_yield_lm)[2]))</pre>
res19_c <- as.numeric(as.character(coef(res19_yield_lm)[1]))</pre>
res19sym_x <- round((-res19_b) / (2*res19_a) , digits = 0)
res19sym_y <- round(res19_a*(res19sym_x^2) + res19_b*res19sym_x + res19_c , digits = 1)
res19_fitted_2 <- (fitted(res19_yield_lm)*23)</pre>
res19_df_2 <- data.frame(suppl_data1_res19$N_level_kgha , res19_fitted_2) #creates dataframe
res19_plot <- ggplot(data = suppl_data1_res19 , aes( x = N_level_kgha , y = PI_N_Uptake)) +
     geom\_point(data = suppl\_data1\_res19 , aes ( x = N\_level\_kgha , y = PI\_N\_Uptake) , color = "#F8766D" ,
     geom\_point(data = suppl\_data1\_res19 , aes ( x = N\_level\_kgha , y = GrainYield\_Mgha*23) , color = "#000 | for the color = 1000 | for the
     geom_line(data = res19_df , aes( x = suppl_data1_res19.N_level_kgha , y = res19_fitted) , color = "#F
     geom_line(data = res19_df_2 , aes( x = suppl_data1_res19.N_level_kgha , y = res19_fitted_2) , color =
     scale_y = continuous(breaks = seq(0, 300, by = 100), sec.axis = sec_axis(~./23, breaks = seq(0, 1))
```

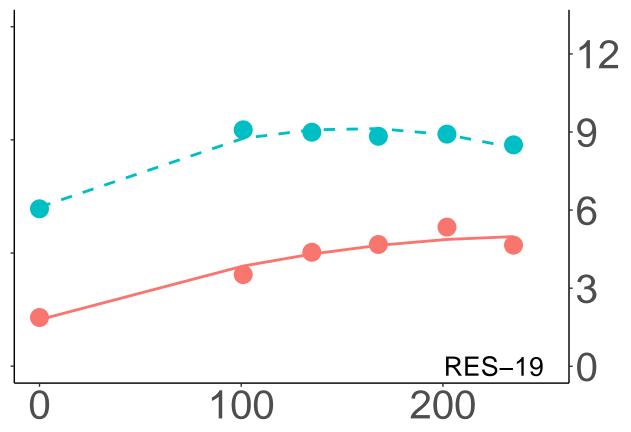
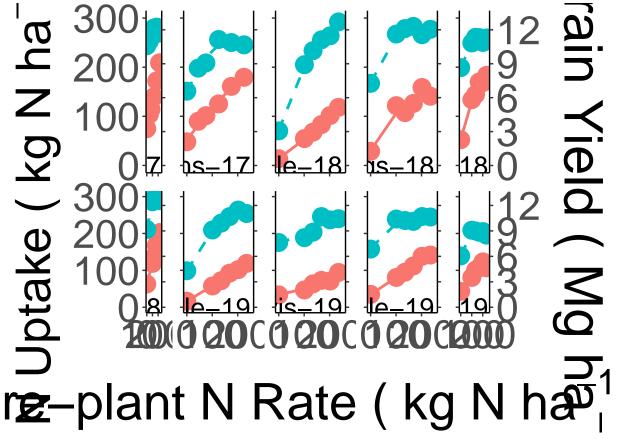


FIGURE 2



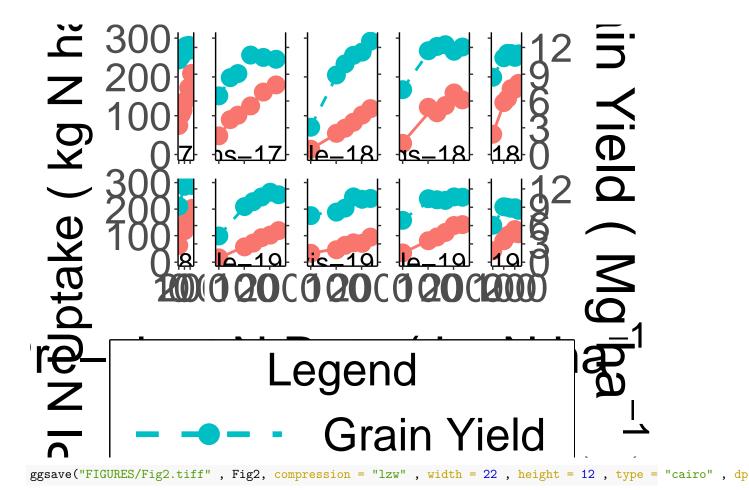


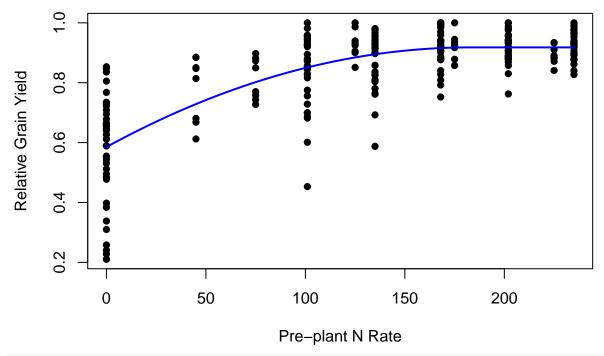
FIGURE S1

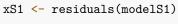
model

```
set.seed(5)
fit.lmS1
            <- lm(relative_grain_yield ~ N_level_kgha , data= paper3_gsdata)</pre>
a.iniS1
            <- fit.lmS1$coefficients[1]</pre>
             <- fit.lmS1$coefficients[2]</pre>
b.iniS1
clx.iniS1
             <- mean(paper3_gsdata$N_level_kgha)</pre>
quadplat <- function(x, a, b, clx) {</pre>
           ifelse(x < clx, a + b * x + (-0.5*b/clx) * x * x,
                              a + b * clx + (-0.5*b/clx) * clx * clx)
modelS1 <- nls(relative_grain_yield ~ quadplat(N_level_kgha, a, b, clx),</pre>
             data = paper3_gsdata,
             start = list(a = a.iniS1,
                          b = b.iniS1,
                          clx = clx.iniS1),
             trace = FALSE,
             nls.control(maxiter = 1000))
```

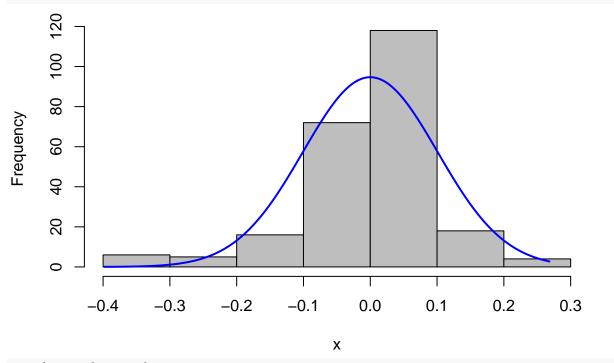
```
summary(modelS1)
## Formula: relative_grain_yield ~ quadplat(N_level_kgha, a, b, clx)
##
## Parameters:
##
        Estimate Std. Error t value Pr(>|t|)
                              37.44
       5.852e-01 1.563e-02
                                     <2e-16 ***
## b 3.633e-03 3.801e-04
                               9.56
                                    <2e-16 ***
## clx 1.832e+02 1.813e+01
                              10.11
                                      <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1011 on 236 degrees of freedom
##
## Number of iterations to convergence: 4
## Achieved convergence tolerance: 4.977e-06
nullfunct <- function(x, m){m}</pre>
m.iniS1
           <- mean(paper3_gsdata$N_level_kgha)</pre>
nullS1 <- nls(relative_grain_yield ~ nullfunct(N_level_kgha, m),</pre>
           data = paper3_gsdata,
           start = list(m = m.iniS1),
           trace = FALSE,
           nls.control(maxiter = 1000))
nagelkerke (modelS1,
           nullS1)
## $Models
## Model: "nls, relative_grain_yield ~ quadplat(N_level_kgha, a, b, clx), paper3_gsdata, list(a = a.ini
## Null: "nls, relative_grain_yield ~ nullfunct(N_level_kgha, m), paper3_gsdata, list(m = m.iniS1), li
## $Pseudo.R.squared.for.model.vs.null
                                Pseudo.R.squared
##
## McFadden
                                       -1.037770
## Cox and Snell (ML)
                                        0.591382
## Nagelkerke (Cragg and Uhler)
                                       -0.432033
## $Likelihood.ratio.test
## Df.diff LogLik.diff Chisq
                                 p.value
##
               -106.95 213.9 3.5679e-47
## $Number.of.observations
## Model: 239
## Null: 239
##
## $Messages
## [1] "Note: For models fit with REML, these statistics are based on refitting with ML"
```

```
## $Warnings
## [1] "None"
confint2(modelS1,
 level = 0.95)
             2.5 %
                         97.5 %
## a 5.543785e-01 6.159608e-01
## b 2.884602e-03 4.382123e-03
## clx 1.474942e+02 2.189150e+02
BootS1 <- nlsBoot(modelS1)</pre>
summary(BootS1)
##
## -----
## Bootstrap statistics
##
          Estimate Std. error
## a 5.855129e-01 1.591681e-02
## b 3.653694e-03 3.970534e-04
## clx 1.844543e+02 1.910856e+01
##
## ----
## Median of bootstrap estimates and percentile confidence intervals
            Median
                           2.5%
                                       97.5%
## a 5.853904e-01 5.543982e-01 6.146719e-01
## b 3.631052e-03 2.939111e-03 4.468332e-03
## clx 1.839725e+02 1.497551e+02 2.251180e+02
plotPredy(data = paper3_gsdata,
              = N_level_kgha,
         x
         y = relative_grain_yield,
         model = modelS1,
         xlab = "Pre-plant N Rate",
         ylab = "Relative Grain Yield")
```

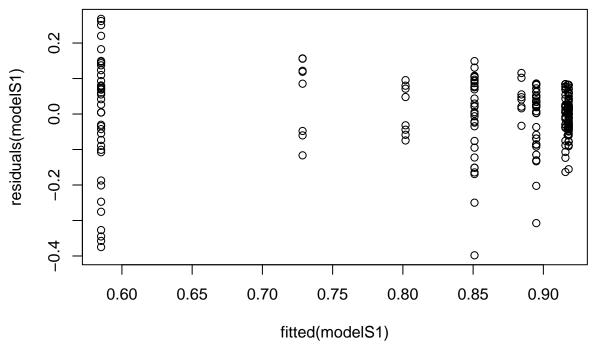




plotNormalHistogram(xS1)



plot(fitted(modelS1),
 residuals(modelS1))



```
aS1 <- summary(modelS1)$coefficients[1]
bS1 <- summary(modelS1)$coefficients[2]
clxS1 <- summary(modelS1)$coefficients[3]
clxS1_SE <- summary(modelS1)$coefficients[6]
clxS1_min <- clxS1 - clxS1_SE
clxS1_max <- clxS1 + clxS1_SE

plateauS1 <- aS1 + bS1 * clxS1 + (-0.5*bS1)*clxS1
plateauS1
```

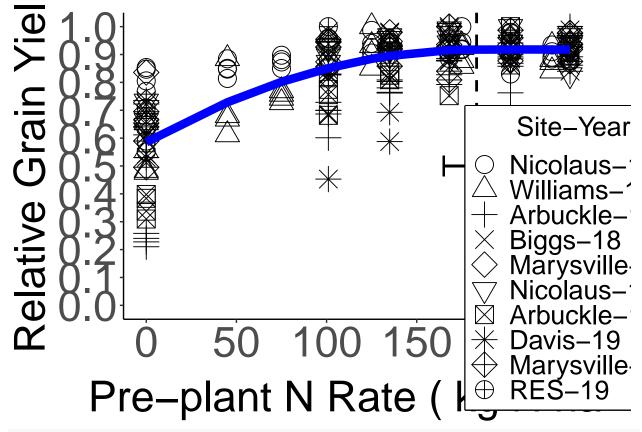
```
## [1] 0.917994
```

```
SErangeS1min <- clxS1 - clxS1_SE
SErangeS1max <- clxS1 + clxS1_SE

Nrate_relyield_fit_quadplat <- fitted(modelS1)
Nrate_relyield_fit_quadplat_df <- data.frame(paper3_gsdata$N_level_kgha , Nrate_relyield_fit_quadplat)
```

plot

```
legend.box.background = element_rect(size = 1)
    ) +
geom_vline(xintercept = 1.832e+02 , linetype = "dashed" , size = 1) +
theme(legend.position = c(0.90 , 0.2)) +
scale_shape_manual(values = c(1:20))
FigS1
```



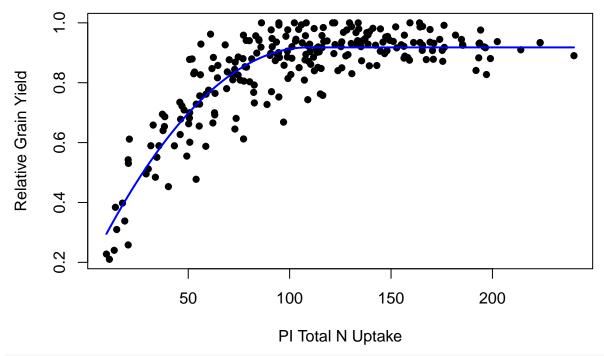
```
ggsave("FIGURES/FigS1.tiff" , FigS1, compression = "lzw" , width = 17 , height = 12 , type = "cairo" ,
```

FIGURE 3

model

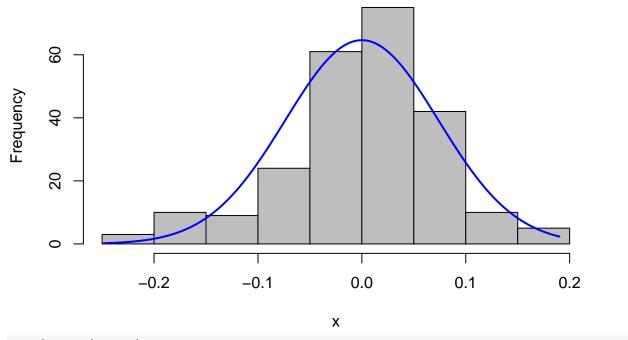
```
data = paper3_gsdata,
            start = list(a = a.ini1,
                         b = b.ini1,
                         clx = clx.ini1),
             trace = FALSE,
             nls.control(maxiter = 1000))
summary(model1)
## Formula: relative_grain_yield ~ quadplat(PI_N_Uptake, a, b, clx)
##
## Parameters:
       Estimate Std. Error t value Pr(>|t|)
##
       1.684e-01 3.206e-02
                             5.252 3.36e-07 ***
## b 1.374e-02 9.857e-04 13.935 < 2e-16 ***
## clx 1.091e+02 4.211e+00 25.919 < 2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.07408 on 236 degrees of freedom
## Number of iterations to convergence: 8
## Achieved convergence tolerance: 2.769e-06
nullfunct <- function(x, m){m}</pre>
m.ini1
         <- mean(paper3_gsdata$PI_N_Uptake)</pre>
null1 <- nls(relative_grain_yield ~ nullfunct(PI_N_Uptake, m),</pre>
           data = paper3_gsdata,
           start = list(m = m.ini1),
           trace = FALSE,
           nls.control(maxiter = 1000))
nagelkerke(model1,
           null1)
## $Models
## Model: "nls, relative_grain_yield ~ quadplat(PI_N_Uptake, a, b, clx), paper3_gsdata, list(a = a.ini1
## Null: "nls, relative_grain_yield ~ nullfunct(PI_N_Uptake, m), paper3_gsdata, list(m = m.ini1), list
## $Pseudo.R.squared.for.model.vs.null
##
                                Pseudo.R.squared
## McFadden
                                       -1.759620
## Cox and Snell (ML)
                                       0.780739
## Nagelkerke (Cragg and Uhler)
                                       -0.570368
## $Likelihood.ratio.test
## Df.diff LogLik.diff Chisq
                                  p.value
##
                -181.34 362.68 1.7572e-79
## $Number.of.observations
##
```

```
## Model: 239
## Null: 239
##
## $Messages
## [1] "Note: For models fit with REML, these statistics are based on refitting with ML"
## $Warnings
## [1] "None"
confint2(model1,
level = 0.95)
##
            2.5 %
                       97.5 %
## a
       0.10521865 0.23154651
## b
       0.01179405 0.01567776
## clx 100.85138096 117.44336986
Boot1 <- nlsBoot(model1)</pre>
summary(Boot1)
## -----
## Bootstrap statistics
        Estimate Std. error
      0.16729414 0.0320060139
## b
      0.01377746 0.0009889959
## clx 109.29780029 4.1986267563
## -----
## Median of bootstrap estimates and percentile confidence intervals
           Median 2.5% 97.5%
       ## a
        0.01368214 0.01197906 0.01585434
## clx 109.33179495 101.43556367 118.04581061
plotPredy(data = paper3_gsdata,
        x = PI_N_Uptake,
         y = relative_grain_yield,
        model = model1,
        xlab = "PI Total N Uptake",
        ylab = "Relative Grain Yield")
```

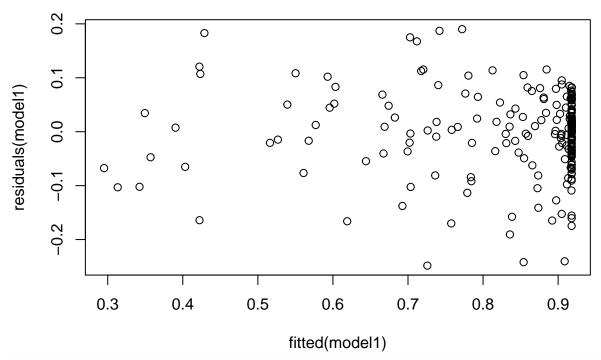


x1 <- residuals(model1)</pre>

plotNormalHistogram(x1)



plot(fitted(model1),
 residuals(model1))



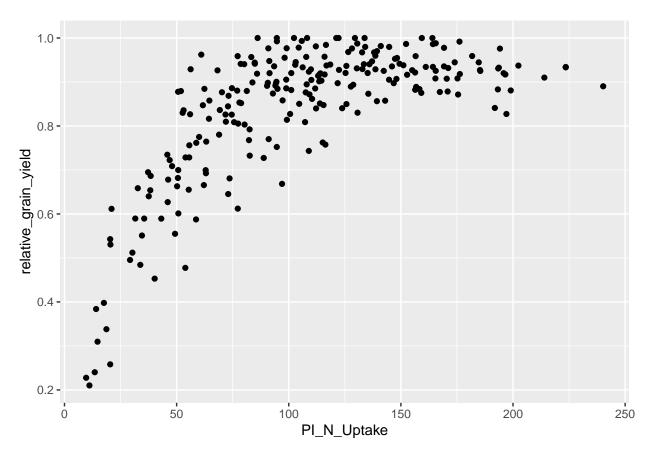
```
a1 <- summary(model1)$coefficients[1]
b1 <- summary(model1)$coefficients[2]
clx1 <- summary(model1)$coefficients[3]
clx1_SE <- summary(model1)$coefficients[6]
clx1_max <- clx1 + clx1_SE
clx1_min <- clx1 - clx1_SE

plateau1 <- a1 + b1 * clx1 + (-0.5*b1)*clx1
plateau1
```

```
## [1] 0.9180015
```

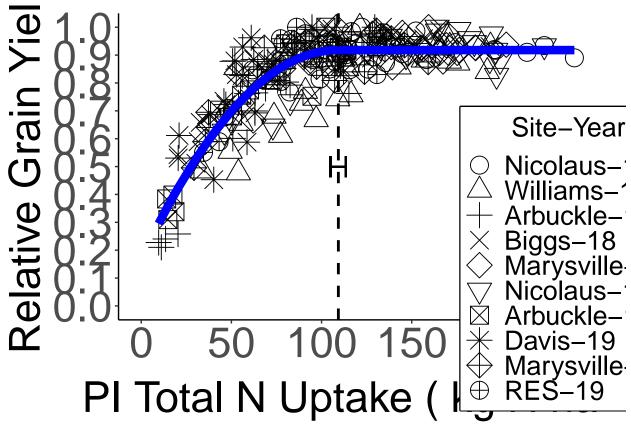
```
pinup_yield_fit_linplat <- fitted(model1)
pinup_yield_fit_linplat_df <- data.frame(paper3_gsdata$PI_N_Uptake , pinup_yield_fit_linplat) #creates

ggplot(data = paper3_gsdata , aes(x = PI_N_Uptake , y = relative_grain_yield)) +
    geom_point(data = paper3_gsdata , aes(x = PI_N_Uptake , y = relative_grain_yield))</pre>
```



plot

```
Fig3 <- ggplot(data = paper3_gsdata , aes ( x = PI_N_Uptake , y = relative_grain_yield)) +
  geom_point(data = paper3_gsdata , aes ( x = PI_N_Uptake , y = relative_grain_yield , shape = site_yea
  geom_line(data = pinup_yield_fit_linplat_df , aes( x = paper3_gsdata.PI_N_Uptake , y = pinup_yield_fi
  coord_cartesian(ylim = c(0, 1), xlim = c(0, 252)) +
  scale_x_continuous(breaks = seq(0, 250, by = 50)) +
  scale_y_continuous(breaks = seq(0, 1, by = .1)) +
  theme_classic() +
  labs(x = "PI Total N Uptake (kg N ha"^-1~")", y = "Relative Grain Yield", shape = "Site-Year") +
  geom\_errorbar(mapping = aes(x = clx1 , y = 0.5 , xmax = clx1_max , xmin = clx1_min) , width = 0.075 ,
  theme(axis.title = element_text(size = 32),
       axis.text = element_text(size = 32),
        legend.text = element_text(size = 20),
        legend.title = element_text(size = 20 , hjust = 0.5),
        legend.box.background = element_rect(size = 1)
        ) +
  geom_vline(xintercept = 109.31388375 , linetype = "dashed" , size = 1) +
  theme(legend.position = c(0.90, 0.2)) +
  scale_shape_manual(values = c(1:20))
Fig3
```



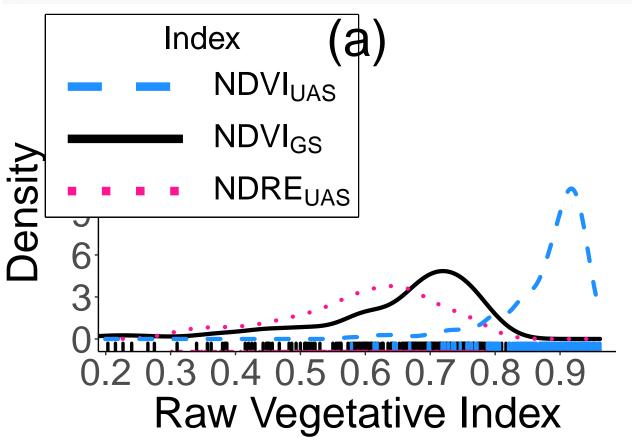
ggsave("FIGURES/Fig3.tiff" , Fig3, compression = "lzw" , width = 17 , height = 12 , type = "cairo" , dp

FIGURE 4

IV Kernel

```
Fig4.1 \leftarrow ggplot(data = paper3_data , aes(x = Index)) +
  geom_density(data = paper3_data, aes(x = Index , color = Platform , linetype = Platform) , size = 1.
  geom_rug(data = paper3_data , aes(x = Index , color = Platform , linetype = Platform) , size = 1.2 ,
  theme_classic() +
  labs(title = "(a)" , x = "Raw Vegetative Index" , y = "Density" , color = "Index" , linetype = "Index
  guides(color = guide_legend(keywidth = 7.5 , keyheight = 1.5 , unit = "cm" , title.hjust = 0.5 , byro
         linetype = guide_legend(keywidth = 7.5 , keyheight = 1.5 , unit = "cm" , override.aes = list(s
  theme(axis.title = element_text(size = 32),
       axis.text = element_text(size = 28),
       legend.text = element_text(size = 24),
        legend.title = element_text(size = 24 , hjust = 0.5),
        legend.position = c(.20, .85),
        legend.text.align = 0,
        legend.box.background = element_rect(size = 1),
        plot.title = element_text(size = 38 , hjust = 0.5)) +
  coord_cartesian(xlim = c(0.225, 0.95), ylim = c(0, 18)) +
  scale_x_continuous(breaks = seq(.2, .9, by = .1)) +
  scale_y\_continuous(breaks = seq(0, 18, by = 3)) +
 scale_color_manual(breaks = c("sUAS NDVI", "GreenSeeker NDVI", "sUAS NDRE"), values = c( "dodgerb
  scale_linetype_manual(breaks = c("sUAS NDVI" , "GreenSeeker NDVI" , "sUAS NDRE") , values = c("dashe
```

Fig4.1



ggsave("FIGURES/NOT_IN_PUBLICATION/Fig4.1.tiff" , Fig4.1 , compression = "lzw" , width = 15 , height =

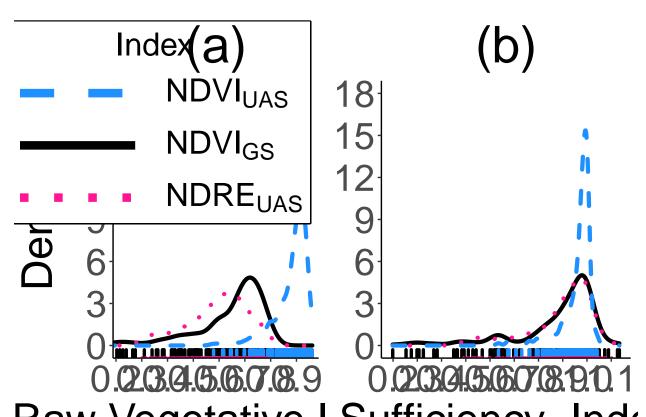
SI Kernel

```
Fig4.2 \leftarrow ggplot(data = paper3_data , aes(x = SI)) +
  geom_density(\frac{data}{data} = paper3_data, aes(x = SI , color = Platform , linetype = Platform) , size = 1.5)
  geom_rug(data = paper3_data , aes(x = SI , color = Platform , linetype = Platform) , size = 1.2) +
  theme_classic() +
  labs(title = "(b)", x = "Sufficiency-Index", y = "Density", color = "Index", linetype = "Index")
  theme(axis.title = element_text(size = 32),
        axis.text = element_text(size = 28),
        legend.text = element_text(size = 24),
        legend.title = element_text(size = 24),
        legend.position = c(0.22, 0.87),
        plot.title = element_text(size = 38 , hjust = 0.5)
        ) +
  coord_cartesian(ylim = c(0, 18)) +
  scale_x_continuous(breaks = seq(.2 , 1.2 , by = .1)) +
  scale_y_continuous(breaks = seq(0 , 18 , by = 3)) +
  scale_color_manual(breaks = c("sUAS NDVI" , "GreenSeeker NDVI" , "sUAS NDRE") , values = c( "dodgerb
  scale_linetype_manual(breaks = c("sUAS NDVI" , "GreenSeeker NDVI" , "sUAS NDRE") , values = c("dashe
    color = guide_legend(byrow = TRUE)
```

```
Fig4.2

Index

I
```



Raw Vegetative I Sufficiency—Indegrave("FIGURES/Fig4.tiff", Fig4, compression = "lzw", width = 20, height = 10, type = "cairo", dp

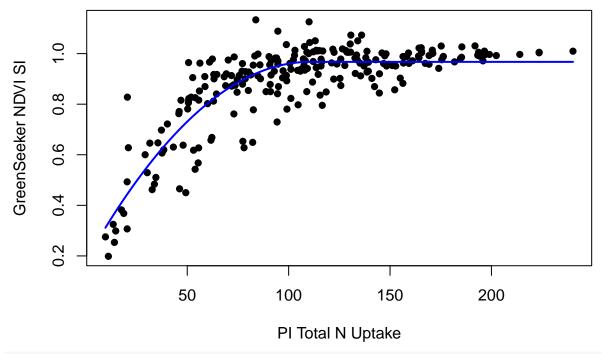
```
FIGURE 5
```

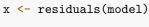
GS-NDVI-SI

```
fit.lm
         <- lm(SI ~ PI_N_Uptake, data= paper3_gsdata)
a.ini
         <- fit.lm$coefficients[1]
b.ini
         <- fit.lm$coefficients[2]</pre>
clx.ini <- mean(paper3_gsdata$PI_N_Uptake)</pre>
quadplat <- function(x, a, b, clx) {</pre>
           ifelse(x < clx, a + b * x)
                                        + (-0.5*b/clx) * x * x,
                             a + b * clx + (-0.5*b/clx) * clx * clx)
model <- nls(SI ~ quadplat(PI_N_Uptake, a, b, clx),</pre>
            data = paper3_gsdata,
            start = list(a = a.ini,
                         b = b.ini,
                          clx = clx.ini),
             trace = FALSE,
             nls.control(maxiter = 1000))
summary(model)
```

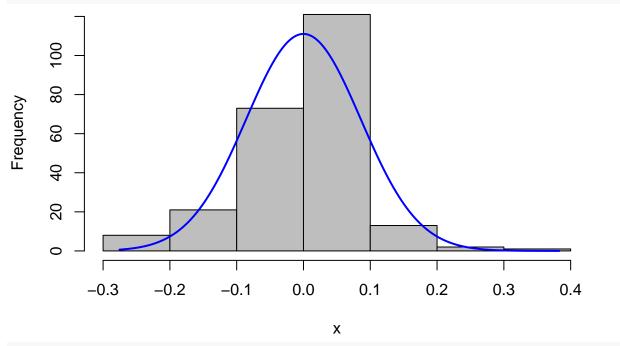
```
model
##
## Formula: SI ~ quadplat(PI_N_Uptake, a, b, clx)
## Parameters:
##
        Estimate Std. Error t value Pr(>|t|)
       1.800e-01 3.691e-02
                             4.876 1.99e-06 ***
       1.424e-02 1.122e-03 12.693 < 2e-16 ***
## clx 1.106e+02 4.709e+00 23.488 < 2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
\#\# Residual standard error: 0.08624 on 236 degrees of freedom
## Number of iterations to convergence: 8
## Achieved convergence tolerance: 4.917e-06
a.ini <- summary(model)$coefficients[1]</pre>
b.ini <- summary(model)$coefficients[2]</pre>
c.ini <- summary(model)$coefficients[3]</pre>
nullfunct <- function(x, m){m}</pre>
         <- mean(paper3_gsdata$SI)</pre>
m.ini
null <- nls(SI ~ nullfunct(PI_N_Uptake, m),</pre>
           data = paper3_gsdata,
           start = list(m = m.ini),
           trace = FALSE,
           nls.control(maxiter = 1000))
nagelkerke(model,
           null)
## $Models
## Model: "nls, SI ~ quadplat(PI_N_Uptake, a, b, clx), paper3_gsdata, list(a = a.ini, b = b.ini, clx =
## Null: "nls, SI ~ nullfunct(PI_N_Uptake, m), paper3_gsdata, list(m = m.ini), list(1000, 1e-05, 0.000
## $Pseudo.R.squared.for.model.vs.null
##
                                Pseudo.R.squared
## McFadden
                                        -1.952810
## Cox and Snell (ML)
                                        0.746651
## Nagelkerke (Cragg and Uhler)
                                        -0.732034
## $Likelihood.ratio.test
## Df.diff LogLik.diff Chisq
                                   p.value
                -164.07 328.14 5.5507e-72
##
         -2
##
## $Number.of.observations
## Model: 239
## Null: 239
```

```
##
## $Messages
## [1] "Note: For models fit with REML, these statistics are based on refitting with ML"
## $Warnings
## [1] "None"
confint2(model,
       level = 0.95)
             2.5 %
                        97.5 %
## a
                    0.25270497
       0.10725897
        0.01202864
                    0.01644837
## clx 101.31668682 119.86910242
Boot <- nlsBoot(model)</pre>
summary(Boot)
##
## -----
## Bootstrap statistics
        Estimate Std. error
##
       0.17888932 0.035694330
## b
       0.01428261 0.001076964
## clx 110.69030394 4.593538755
## -----
## Median of bootstrap estimates and percentile confidence intervals
            Median
                         2.5%
                                    97.5%
        0.18009300 0.11137336 0.24562942
## a
        ## clx 110.54958982 102.01768054 119.99260674
plotPredy(data = paper3_gsdata,
         x = PI_N_Uptake,
         У
             = SI,
         model = model,
         xlab = "PI Total N Uptake",
         ylab = "GreenSeeker NDVI SI")
```





plotNormalHistogram(x)



```
0.4
                                 0
       3
       o.
                                                                       0
      0.2
residuals(model)
                                  0
      0.1
                                 0
                            00
                                               0
                0
                      0
      -0.1
                        0
                                                  00
                                                                    0
                       0
                  0
                                 0
                                                                         00
                                                                          00
                                                                  0
                                                                                            80
                                                                      0
             0.3
                          0.4
                                       0.5
                                                                              8.0
                                                    0.6
                                                                 0.7
                                                                                           0.9
                                                   fitted(model)
a <- summary(model)$coefficients[1]</pre>
```

```
b <- summary(model)$coefficients[2]
clx <- summary(model)$coefficients[3]

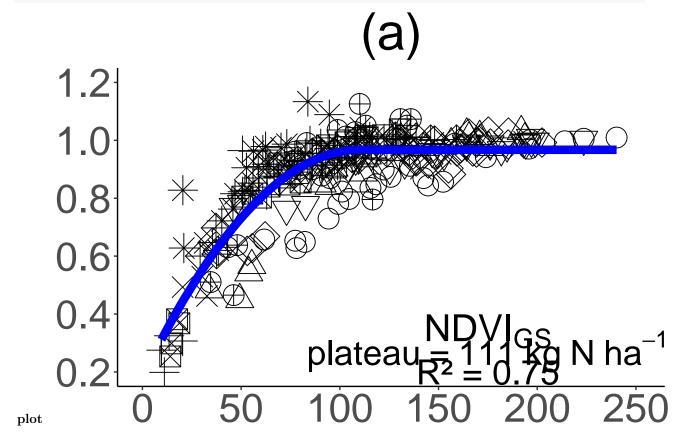
plateau <- a + b * clx + (-0.5*b)*clx
round(plateau , digits = 2)

## [1] 0.97</pre>
```

GS_ndvi_fit_quadplat <- fitted(model)
GS_ndvi_quadplat_df <- data.frame(paper3_gsdata\$PI_N_Uptake , GS_ndvi_fit_quadplat) #creates dataframe

```
GS_ndvi_si_quad_plot \leftarrow ggplot(data = paper3_gsdata , aes ( x = PI_N_Uptake , y = SI)) +
  geom_point(data = paper3_gsdata , aes ( x = PI_N_Uptake , y = SI , shape = site_year) , size = 7) +
  geom\_line(data = GS\_ndvi\_quadplat\_df , aes(x = paper3\_gsdata.PI\_N\_Uptake , y = GS\_ndvi\_fit\_quadplat)
  coord_cartesian(ylim = c(0.20 , 1.2) , xlim = c(0 , 252)) +
  scale_x_continuous(breaks = seq(0, 250, by = 50)) +
  scale_y_continuous(breaks = seq(.2 , 1.2 , by = .2)) +
  theme classic() +
  labs(title = "(a)", x = NULL, y = NULL) +
  annotate("text", x = 175, y = .34, label = "NDVI[GS]", size = 10, color = "black", hjust = .5, pa
  annotate("text", x = 175, y = .27, label = "plateau~'='~111~kg~N~ha^-1", size = 9, color = "black"
  annotate("text", x = 175, y = .2, label = "R^2 = 0.75", size = 9, color = "black", hjust = .5) +
  theme(axis.title = element_text(size = 30),
       axis.text = element_text(size = 30),
        legend.text = element_text(size = 24),
        legend.title = element_text(size = 24),
        plot.title = element_text(size = 38 , hjust = 0.5)) +
  geom_hline(yintercept = 0 , size = 0.75) +
  theme(legend.position = "none") +
  scale_shape_manual(values = c(1:20))
```

GS_ndvi_si_quad_plot



UAS-NDRE-SI

```
fit.lm2
         <- lm(SI ~ PI_N_Uptake, data= paper3_uas_ndre_data)</pre>
a.ini2
          <- fit.lm2$coefficients[1]</pre>
           <- fit.lm2$coefficients[2]</pre>
b.ini2
           <- mean(paper3_uas_ndre_data$PI_N_Uptake)</pre>
clx.ini2
quadplat <- function(x, a, b, clx) {</pre>
           ifelse(x < clx, a + b * x + (-0.5*b/clx) * x * x,
                             a + b * clx + (-0.5*b/clx) * clx * clx)
model2 <- nls(SI ~ quadplat(PI_N_Uptake, a, b, clx),</pre>
            data = paper3_uas_ndre_data,
            start = list(a = a.ini2,
                         b = b.ini2,
                          clx = clx.ini2),
             trace = FALSE,
             nls.control(maxiter = 1000))
summary(model2)
```

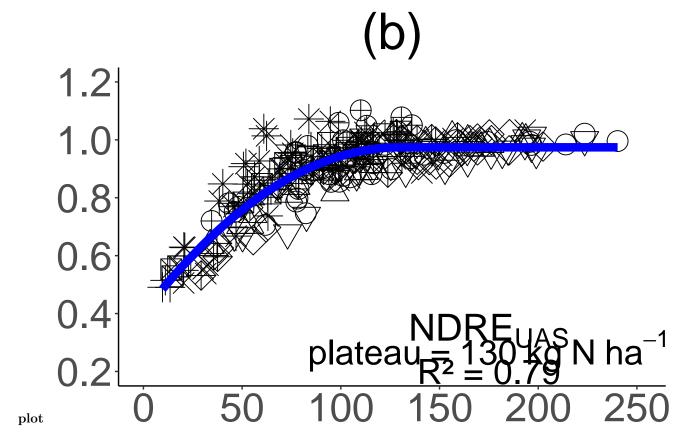
model

##

```
## Formula: SI ~ quadplat(PI_N_Uptake, a, b, clx)
##
## Parameters:
       Estimate Std. Error t value Pr(>|t|)
##
## a
      4.025e-01 2.270e-02
                              17.73 <2e-16 ***
## b 8.801e-03 5.988e-04
                              14.70
                                    <2e-16 ***
## clx 1.299e+02 5.146e+00
                              25.25
                                      <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.06073 on 236 degrees of freedom
## Number of iterations to convergence: 7
## Achieved convergence tolerance: 1.431e-06
nullfunct <- function(x, m){m}
m.ini2
          <- mean(paper3_uas_ndre_data$SI)</pre>
null2 <- nls(SI ~ nullfunct(PI_N_Uptake, m),</pre>
           data = paper3_uas_ndre_data,
           start = list(m = m.ini2),
           trace = FALSE,
           nls.control(maxiter = 1000))
nagelkerke(model2,
          nul12)
## $Models
##
## Model: "nls, SI ~ quadplat(PI_N_Uptake, a, b, clx), paper3_uas_ndre_data, list(a = a.ini2, b = b.ini
## Null: "nls, SI ~ nullfunct(PI_N_Uptake, m), paper3_uas_ndre_data, list(m = m.ini2), list(1000, 1e-0
## $Pseudo.R.squared.for.model.vs.null
                                Pseudo.R.squared
## McFadden
                                       -1.277960
## Cox and Snell (ML)
                                        0.789474
## Nagelkerke (Cragg and Uhler)
                                       -0.331067
## $Likelihood.ratio.test
## Df.diff LogLik.diff Chisq
                                 p.value
##
         -2
                -186.2 372.4 1.3644e-81
##
## $Number.of.observations
##
## Model: 239
## Null: 239
##
## $Messages
## [1] "Note: For models fit with REML, these statistics are based on refitting with ML"
## $Warnings
## [1] "None"
```

```
confint2(model2,
     level = 0.95)
##
              2.5 %
                          97.5 %
## a
       3.578209e-01
                      0.44726792
## b
      7.621308e-03
                      0.00998052
## clx 1.197706e+02 140.04619274
Boot2 <- nlsBoot(model2)</pre>
summary(Boot2)
##
## -----
## Bootstrap statistics
           Estimate
                      Std. error
     4.011069e-01 0.0226212776
## a
## b 8.847397e-03 0.0006002618
## clx 1.298601e+02 5.1079241184
##
## -----
## Median of bootstrap estimates and percentile confidence intervals
                            2.5%
                                         97.5%
             Median
       4.013248e-01 3.570472e-01
                                    0.44539348
## a
       8.830578e-03 7.718934e-03
                                  0.01003484
## clx 1.296631e+02 1.201484e+02 141.16991967
plotPredy(data = paper3_uas_ndre_data,
               = PI_N_Uptake,
          X
               = SI,
          model = model2,
          xlab = "PI Total N Uptake",
          ylab = "sUAS NDRE SI")
      1.0
      0.9
SUAS NDRE SI
      0.8
      0.7
      9.0
      S
                         50
                                       100
                                                      150
                                                                     200
                                       PI Total N Uptake
```

```
a2 <- summary(model2)$coefficients[1]</pre>
b2 <- summary(model2)$coefficients[2]
clx2 <- summary(model2)$coefficients[3]</pre>
plateau2 <- a2 + b2 * clx2 + (-0.5*b2)*clx2
round(plateau2 , digits = 2)
## [1] 0.97
min(paper3_uas_ndre_data$SI)
## [1] 0.4903767
sUAS_ndre_fit_quadplat <- fitted(model2)</pre>
sUAS_ndre_quadplat_df <- data.frame(paper3_uas_ndre_data$PI_N_Uptake , sUAS_ndre_fit_quadplat) #creates
suas_ndre_si_quad_plot <- ggplot(data = paper3_uas_ndre_data , aes ( x = PI_N_Uptake , y = SI)) +</pre>
      geom_point(data = paper3_uas_ndre_data , aes ( x = PI_N_Uptake , y = SI , shape = site_year) , size =
      \texttt{geom\_line}(\texttt{data} = \texttt{sUAS\_ndre\_quadplat\_df} \text{ , aes}( \text{ } \textbf{x} = \texttt{paper3\_uas\_ndre\_data.PI\_N\_Uptake} \text{ , } \textbf{y} = \texttt{sUAS\_ndre\_fire}(\texttt{supersolution}) \text{ } \textbf{y} = \texttt{supersolution}(\texttt{value}) \text{ } \textbf{y} = \texttt{s
      coord_cartesian(ylim = c(0.20, 1.2), xlim = c(0, 252)) +
      scale_x_continuous(breaks = seq(0, 250, by = 50)) +
      scale_y_continuous(breaks = seq(.2 , 1.2 , by = .2)) +
      theme_classic() +
      labs(title = "(b)", x = NULL, y = NULL) +
      annotate("text", x = 175, y = .34, label = "NDRE[UAS]", size = 10, color = "black", hjust = .5, p
      annotate("text", x = 175, y = .27, label = "plateau~'='~130~kg~N~ha^-1", size = 9, color = "black"
      annotate("text", x = 175, y = .2, label = "R^2 = 0.79", size = 9, color = "black", hjust = .5) +
      theme(axis.title = element_text(size = 30),
                       axis.text = element_text(size = 30),
                       legend.text = element_text(size = 24),
                       legend.title = element_text(size = 24),
                       plot.title = element_text(size = 38 , hjust = 0.5)) +
      geom_hline(yintercept = 0 , size = 0.75) +
      theme(legend.position = "none") +
      scale_shape_manual(values = c(1:20))
suas_ndre_si_quad_plot
```



UAS-NDVI-SI

```
fit.lm3
           <- lm(SI ~ PI_N_Uptake, data= paper3_uas_ndvi_data)</pre>
a.ini3
         <- fit.lm3$coefficients[1]</pre>
b.ini3
          <- fit.lm3$coefficients[2]</pre>
clx.ini3 <- mean(paper3_uas_ndvi_data$PI_N_Uptake)</pre>
quadplat <- function(x, a, b, clx) {</pre>
           ifelse(x < clx, a + b * x + (-0.5*b/clx) * x * x,
                             a + b * clx + (-0.5*b/clx) * clx * clx)
model3 <- nls(SI ~ quadplat(PI_N_Uptake, a, b, clx),</pre>
            data = paper3_uas_ndvi_data,
            start = list(a = a.ini3,
                          b = b.ini3,
                          clx = clx.ini3),
             trace = FALSE,
             nls.control(maxiter = 1000))
summary(model3)
```

model

```
##
## Formula: SI ~ quadplat(PI_N_Uptake, a, b, clx)
##
```

```
## Parameters:
##
       Estimate Std. Error t value Pr(>|t|)
      6.376e-01 1.359e-02
                            46.92 <2e-16 ***
## b 7.317e-03 4.689e-04
                              15.61
                                      <2e-16 ***
## clx 9.622e+01 3.193e+00
                              30.13
                                    <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.02818 on 236 degrees of freedom
##
## Number of iterations to convergence: 11
## Achieved convergence tolerance: 6.253e-06
nullfunct <- function(x, m){m}</pre>
m.ini3
          <- mean(paper3_uas_ndvi_data$SI)</pre>
null3 <- nls(SI ~ nullfunct(PI_N_Uptake, m),</pre>
           data = paper3_uas_ndvi_data,
           start = list(m = m.ini3),
           trace = FALSE,
           nls.control(maxiter = 1000))
nagelkerke(model3,
           nul13)
## $Models
## Model: "nls, SI ~ quadplat(PI_N_Uptake, a, b, clx), paper3_uas_ndvi_data, list(a = a.ini3, b = b.ini
## Null: "nls, SI ~ nullfunct(PI_N_Uptake, m), paper3_uas_ndvi_data, list(m = m.ini3), list(1000, 1e-0
## $Pseudo.R.squared.for.model.vs.null
##
                                Pseudo.R.squared
## McFadden
                                      -0.6636080
## Cox and Snell (ML)
                                       0.8210270
## Nagelkerke (Cragg and Uhler)
                                      -0.0663964
## $Likelihood.ratio.test
## Df.diff LogLik.diff Chisq
                                 p.value
##
                -205.6 411.2 5.1057e-90
## $Number.of.observations
## Model: 239
## Null: 239
##
## $Messages
## [1] "Note: For models fit with REML, these statistics are based on refitting with ML"
## $Warnings
## [1] "None"
confint2(model3,
       level = 0.95)
##
              2.5 %
                          97.5 %
```

```
0.610795961
                     0.66433966
       0.006393222
                     0.00824059
## clx 89.926042684 102.50808027
Boot3 <- nlsBoot(model3)</pre>
summary(Boot3)
##
## -----
## Bootstrap statistics
##
          Estimate
                     Std. error
       0.637933201 0.0137008534
       0.007308095 0.0004632789
## b
## clx 96.422242952 3.1173968820
##
##
## Median of bootstrap estimates and percentile confidence intervals
            Median
                          2.5%
                                      97.5%
       ## a
       ## clx 96.377448182 90.447467763 1.024119e+02
plotPredy(data = paper3_uas_ndvi_data,
              = PI_N_Uptake,
         X
         У
              = SI,
         model = model3,
         xlab = "PI Total N Uptake",
         ylab = "sUAS NDVI SI")
      1.0
SUAS NDVI SI
     6.0
     0.8
                       50
                                     100
                                                   150
                                                                 200
                                     PI Total N Uptake
a3 <- summary(model3)$coefficients[1]</pre>
b3 <- summary(model3)$coefficients[2]</pre>
clx3 <- summary(model3)$coefficients[3]</pre>
```

```
plateau3 <- a3 + b3 * clx3 + (-0.5*b3)*clx3
round(plateau3 , digits = 2)
## [1] 0.99
min(paper3_uas_ndvi_data$SI)
## [1] 0.6349517
sUAS_ndvi_fit_quadplat <- fitted(model3)</pre>
sUAS_ndvi_quadplat_df <- data.frame(paper3_uas_ndvi_data$PI_N_Uptake , sUAS_ndvi_fit_quadplat) #creates
suas_ndvi_si_quad_plot \leftarrow ggplot(data = paper3_uas_ndvi_data , aes ( x = PI_N_Uptake , y = SI)) +
  geom_point(data = paper3_uas_ndvi_data , aes ( x = PI_N_Uptake , y = SI , shape = site_year) , size =
  geom_line(data = sUAS_ndvi_quadplat_df , aes( x = paper3_uas_ndvi_data.PI_N_Uptake , y = sUAS_ndvi_fi
  coord_cartesian(ylim = c(0.20, 1.2), xlim = c(0, 252)) +
  scale_x_continuous(breaks = seq(0, 250, by = 50)) +
  scale_y\_continuous(breaks = seq(.2 , 1.2 , by = .2)) +
  theme_classic() +
  labs(title = "(c)" , x = NULL , y = NULL , shape = "Site-Year") +
  annotate("text", x = 175, y = .34, label = "NDVI[UAS]", size = 10, color = "black", hjust = .5, p
  annotate("text", x = 175, y = .27, label = "plateau~'='~96~kg~N~ha^-1", size = 9, color = "black",
  annotate("text", x = 175, y = .2, label = "R^2 = 0.82", size = 9, color = "black", hjust = .5) +
  theme(axis.title = element_text(size = 30),
       axis.text = element_text(size = 30),
       legend.text = element_text(size = 18),
       legend.title = element_text(size = 18 , hjust = 0.5),
       legend.box.background = element_rect(size = 1),
        plot.title = element_text(size = 38 , hjust = 0.5)) +
  geom_hline(yintercept = 0 , size = 0.75) +
  theme(legend.position = c(.66, .425)) +
  scale_shape_manual(values = c(1:20)) +
  guides(shape = guide_legend(override.aes = list(size=4)))
suas_ndvi_si_quad_plot
```

(c)

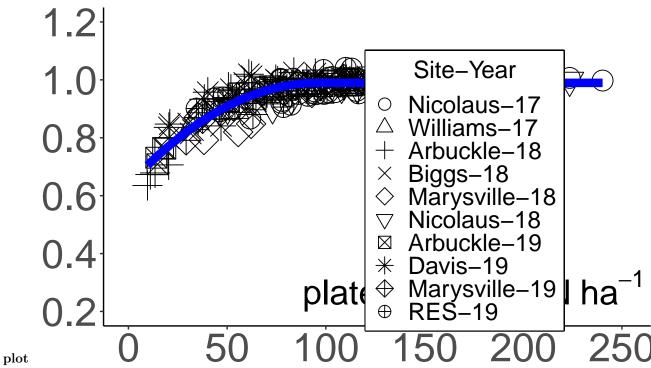
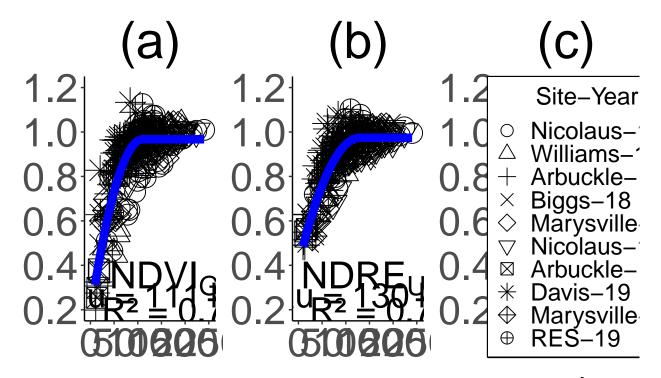
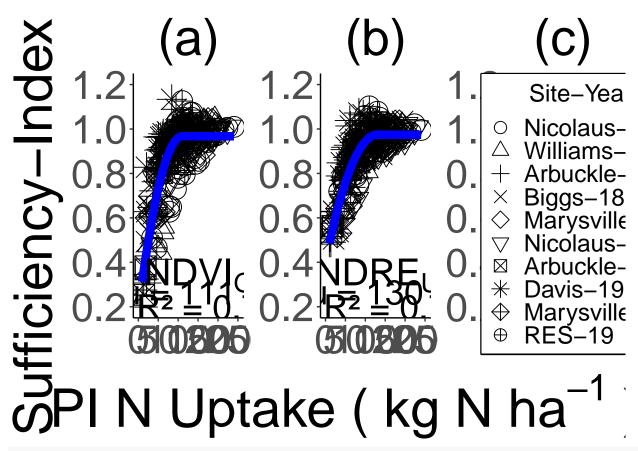


FIGURE 5



PI N Uptake (kg N ha⁻¹)



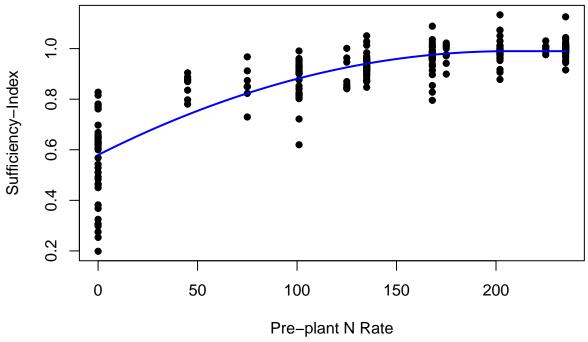
```
ggsave("FIGURES/Fig5.tiff" , Fig5 , compression = "lzw" , width = 22 , height = 10, type = "cairo" , dp
```

model gs NDVI

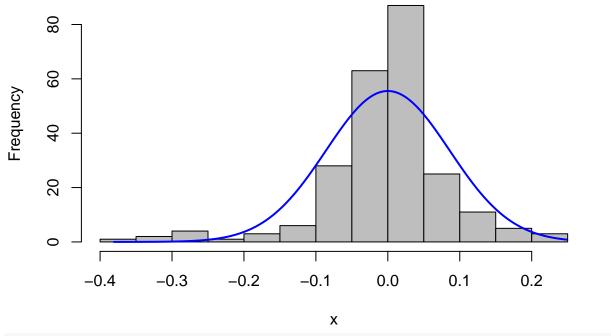
```
set.seed(5)
fit.lmS3_1
              <- lm(SI ~ N_level_kgha , data = paper3_gsdata)</pre>
a.iniS3_1
              <- fit.lmS3_1$coefficients[1]</pre>
b.iniS3_1
              <- fit.lmS3_1$coefficients[2]</pre>
clx.iniS3_1 <- mean(paper3_gsdata$N_level_kgha)</pre>
quadplat <- function(x, a, b, clx) {</pre>
           ifelse(x < clx, a + b * x + (-0.5*b/clx) * x * x,
                             a + b * clx + (-0.5*b/clx) * clx * clx)
modelS3_1 <- nls(SI ~ quadplat(N_level_kgha, a, b, clx),</pre>
            data = paper3_gsdata,
            start = list(a = a.iniS3_1,
                          b = b.iniS3_1,
                          clx = clx.iniS3_1),
             trace = FALSE,
             nls.control(maxiter = 1000))
summary(modelS3_1)
```

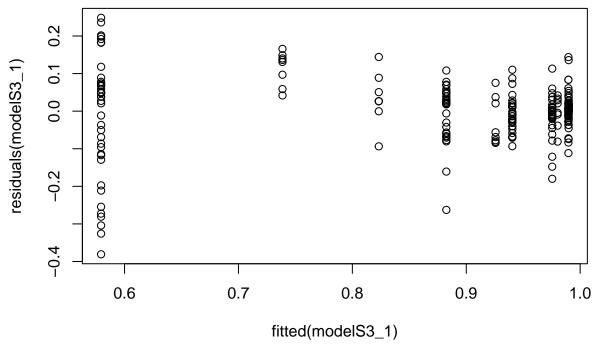
```
##
## Formula: SI ~ quadplat(N_level_kgha, a, b, clx)
## Parameters:
##
        Estimate Std. Error t value Pr(>|t|)
       5.794e-01 1.322e-02
                              43.81
                                      <2e-16 ***
       3.971e-03 2.754e-04
                              14.42
                                      <2e-16 ***
## clx 2.067e+02 1.385e+01
                              14.93
                                     <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
\#\# Residual standard error: 0.08622 on 236 degrees of freedom
## Number of iterations to convergence: 7
## Achieved convergence tolerance: 8.591e-06
nullfunct <- function(x, m){m}</pre>
             <- mean(paper3_gsdata$N_level_kgha)</pre>
m.iniS3_1
nullS3_1 <- nls(SI ~ nullfunct(N_level_kgha, m),</pre>
           data = paper3_gsdata,
           start = list(m = m.iniS3_1),
           trace = FALSE,
           nls.control(maxiter = 1000))
nagelkerke(modelS3_1,
           nullS3 1)
## $Models
## Model: "nls, SI ~ quadplat(N_level_kgha, a, b, clx), paper3_gsdata, list(a = a.iniS3_1, b = b.iniS3_
## Null: "nls, SI ~ nullfunct(N_level_kgha, m), paper3_gsdata, list(m = m.iniS3_1), list(1000, 1e-05,
## $Pseudo.R.squared.for.model.vs.null
##
                                Pseudo.R.squared
## McFadden
                                       -1.953300
## Cox and Snell (ML)
                                        0.746738
## Nagelkerke (Cragg and Uhler)
                                       -0.732118
##
## $Likelihood.ratio.test
## Df.diff LogLik.diff Chisq
                                  p.value
                -164.11 328.23 5.3285e-72
##
         -2
##
## $Number.of.observations
## Model: 239
## Null: 239
##
## $Messages
## [1] "Note: For models fit with REML, these statistics are based on refitting with ML"
## $Warnings
```

```
## [1] "None"
confint2(modelS3_1,
level = 0.95)
##
                        97.5 %
             2.5 %
## a 5.533462e-01 6.054522e-01
## b 3.428442e-03 4.513665e-03
## clx 1.794695e+02 2.340266e+02
BootS3_1 <- nlsBoot(modelS3_1)</pre>
summary(BootS3_1)
## ----
## Bootstrap statistics
## Estimate Std. error
## a 5.798424e-01 1.322227e-02
## b 3.971962e-03 2.840432e-04
## clx 2.078307e+02 1.496834e+01
##
## -----
## Median of bootstrap estimates and percentile confidence intervals
           Median 2.5% 97.5%
## a 5.799457e-01 5.535012e-01 6.042564e-01
## b 3.960685e-03 3.425772e-03 4.547434e-03
## clx 2.072158e+02 1.798692e+02 2.382311e+02
plotPredy(data = paper3_gsdata,
         x = N_level_kgha,
y = SI,
         model = modelS3_1,
         xlab = "Pre-plant N Rate",
         ylab = "Sufficiency-Index")
```



xS3_1 <- residuals(modelS3_1)
plotNormalHistogram(xS3_1)</pre>





```
aS3_1 <- summary(modelS3_1)$coefficients[1]
bS3_1 <- summary(modelS3_1)$coefficients[2]
clxS3_1 <- summary(modelS3_1)$coefficients[3]
clxS3_1_SE <- summary(modelS3_1)$coefficients[6]

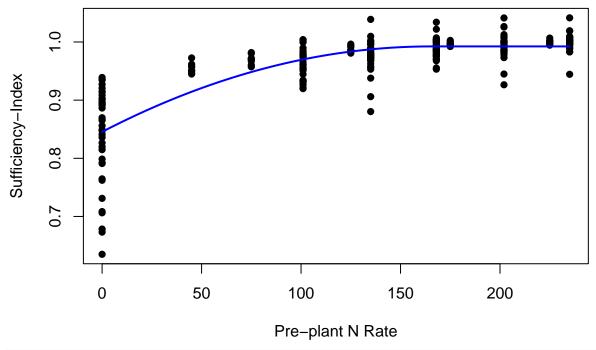
plateauS3_1 <- aS3_1 + bS3_1 * clxS3_1 + (-0.5*bS3_1)*clxS3_1
plateauS3_1
```

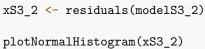
```
## [1] 0.989903
```

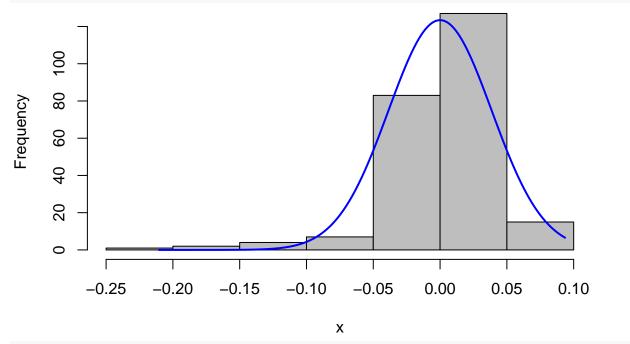
model uas NDVI

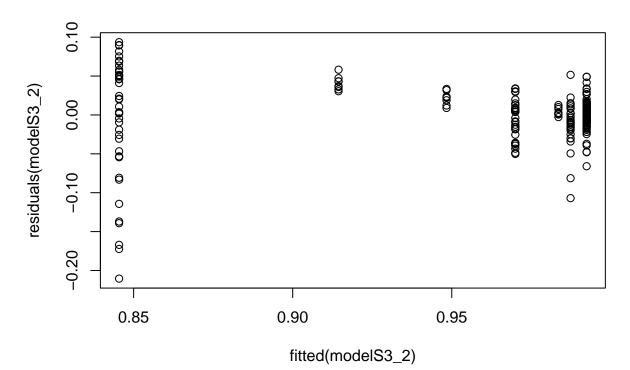
```
modelS3_2 <- nls(SI ~ quadplat(N_level_kgha, a, b, clx),</pre>
            data = paper3_uas_ndvi_data,
            start = list(a = a.iniS3_2,
                        b = b.iniS3_2,
                         clx = clx.iniS3_2),
             trace = FALSE,
             nls.control(maxiter = 1000))
summary(modelS3_2)
##
## Formula: SI ~ quadplat(N_level_kgha, a, b, clx)
##
## Parameters:
##
       Estimate Std. Error t value Pr(>|t|)
      8.454e-01 6.022e-03 140.38 <2e-16 ***
      1.774e-03 1.670e-04
                              10.62
                                    <2e-16 ***
## clx 1.657e+02 1.446e+01
                                     <2e-16 ***
                              11.46
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.03879 on 236 degrees of freedom
##
## Number of iterations to convergence: 10
## Achieved convergence tolerance: 6.398e-06
nullfunct <- function(x, m){m}</pre>
m.iniS3 2
             <- mean(paper3_uas_ndvi_data$N_level_kgha)</pre>
nullS3_2 <- nls(SI ~ nullfunct(N_level_kgha, m),</pre>
           data = paper3_uas_ndvi_data,
           start = list(m = m.iniS3 2),
           trace = FALSE,
           nls.control(maxiter = 1000))
nagelkerke(modelS3_2,
           nullS3_2)
## $Models
## Model: "nls, SI ~ quadplat(N_level_kgha, a, b, clx), paper3_uas_ndvi_data, list(a = a.iniS3_2, b = b
## Null: "nls, SI ~ nullfunct(N_level_kgha, m), paper3_uas_ndvi_data, list(m = m.iniS3_2), list(1000,
## $Pseudo.R.squared.for.model.vs.null
##
                                Pseudo.R.squared
## McFadden
                                      -0.4169650
## Cox and Snell (ML)
                                       0.6607630
## Nagelkerke (Cragg and Uhler)
                                      -0.0534359
##
## $Likelihood.ratio.test
## Df.diff LogLik.diff Chisq p.value
##
              -129.19 258.37 7.854e-57
##
```

```
## $Number.of.observations
##
## Model: 239
## Null: 239
## $Messages
## [1] "Note: For models fit with REML, these statistics are based on refitting with ML"
## $Warnings
## [1] "None"
confint2(modelS3_2,
      level = 0.95)
##
             2.5 %
                         97.5 %
## a 8.335599e-01 8.572886e-01
## b 1.444791e-03 2.102728e-03
## clx 1.372251e+02 1.942110e+02
BootS3_2 <- nlsBoot(modelS3_2)</pre>
summary(BootS3_2)
##
## ----
## Bootstrap statistics
          Estimate
                    Std. error
## a 8.453685e-01 6.051449e-03
## b 1.793578e-03 1.807741e-04
## clx 1.654646e+02 1.530188e+01
##
## -----
## Median of bootstrap estimates and percentile confidence intervals
                    2.5% 97.5%
           Median
## a 8.457973e-01 8.331825e-01 8.562262e-01
## b 1.780891e-03 1.465172e-03 2.178772e-03
## clx 1.652116e+02 1.362356e+02 1.961071e+02
plotPredy(data = paper3_uas_ndvi_data,
         x = N_level_kgha,
         y = SI,
         model = modelS3_2,
         xlab = "Pre-plant N Rate",
         ylab = "Sufficiency-Index")
```









```
aS3_2 <- summary(modelS3_2)$coefficients[1]
bS3_2 <- summary(modelS3_2)$coefficients[2]
clxS3_2 <- summary(modelS3_2)$coefficients[3]
clxS3_2_SE <- summary(modelS3_2)$coefficients[6]

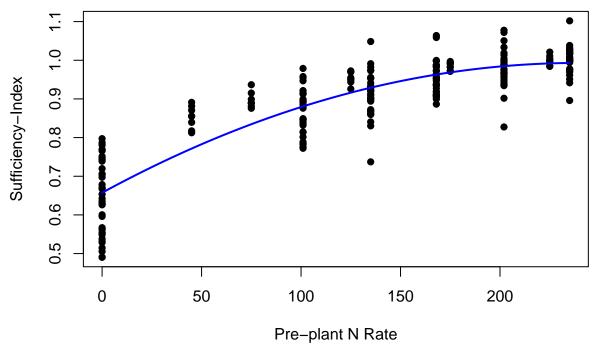
plateauS3_2 <- aS3_2 + bS3_2 * clxS3_2 + (-0.5*bS3_2)*clxS3_2
plateauS3_2
```

```
## [1] 0.9923962
```

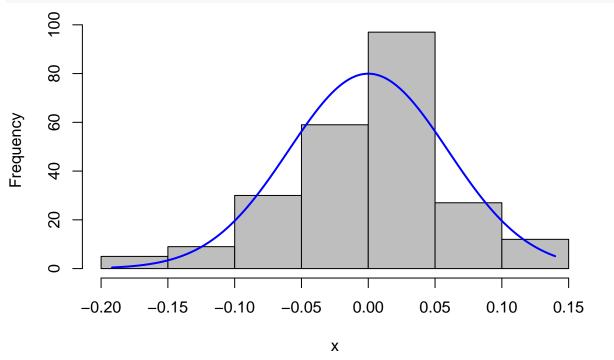
model uas NDRE

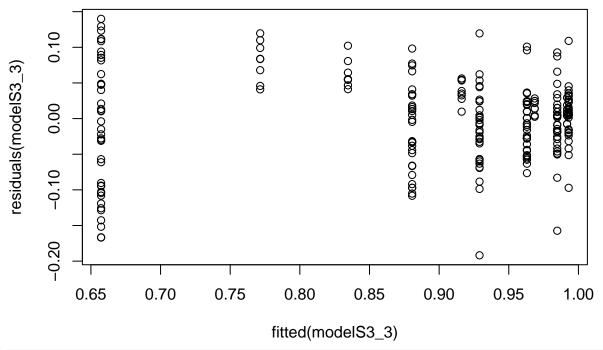
```
modelS3_3 <- nls(SI ~ quadplat(N_level_kgha, a, b, clx),</pre>
            data = paper3_uas_ndre_data,
            start = list(a = a.iniS3_3,
                         b = b.iniS3 3,
                         clx = clx.iniS3_3),
             trace = FALSE,
             nls.control(maxiter = 1000))
summary(modelS3_3)
## Formula: SI ~ quadplat(N_level_kgha, a, b, clx)
##
## Parameters:
##
       Estimate Std. Error t value Pr(>|t|)
## a 6.573e-01 9.098e-03 72.25 <2e-16 ***
## b 2.801e-03 1.667e-04
                              16.81 <2e-16 ***
## clx 2.398e+02 1.529e+01
                             15.69 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.05988 on 236 degrees of freedom
## Number of iterations to convergence: 4
## Achieved convergence tolerance: 5.157e-09
nullfunct <- function(x, m){m}</pre>
m.iniS3 3
            <- mean(paper3_uas_ndre_data$N_level_kgha)</pre>
nullS3_3 <- nls(SI ~ nullfunct(N_level_kgha, m),</pre>
           data = paper3_uas_ndre_data,
           start = list(m = m.iniS3_3),
           trace = FALSE,
           nls.control(maxiter = 1000))
nagelkerke(modelS3_3,
           nullS3_3)
## $Models
##
## Model: "nls, SI ~ quadplat(N_level_kgha, a, b, clx), paper3_uas_ndre_data, list(a = a.iniS3_3, b = b
## Null: "nls, SI ~ nullfunct(N_level_kgha, m), paper3_uas_ndre_data, list(m = m.iniS3_3), list(1000,
## $Pseudo.R.squared.for.model.vs.null
                                Pseudo.R.squared
##
## McFadden
                                       -1.301080
## Cox and Snell (ML)
                                        0.795325
## Nagelkerke (Cragg and Uhler)
                                      -0.333521
## $Likelihood.ratio.test
## Df.diff LogLik.diff Chisq
                                  p.value
##
        -2
              -189.57 379.13 4.7019e-83
##
```

```
## $Number.of.observations
##
## Model: 239
## Null: 239
## $Messages
## [1] "Note: For models fit with REML, these statistics are based on refitting with ML"
## $Warnings
## [1] "None"
confint2(modelS3_3,
      level = 0.95)
##
             2.5 %
                         97.5 %
## a 6.393884e-01 6.752354e-01
## b 2.472646e-03 3.129283e-03
## clx 2.096701e+02 2.699016e+02
BootS3_3 <- nlsBoot(modelS3_3)</pre>
summary(BootS3_3)
##
## ----
## Bootstrap statistics
          Estimate Std. error
## a 6.574265e-01 9.371511e-03
## b 2.801281e-03 1.743101e-04
## clx 2.414021e+02 1.648056e+01
##
## -----
## Median of bootstrap estimates and percentile confidence intervals
                    2.5% 97.5%
           Median
## a 6.577183e-01 6.397810e-01 6.745054e-01
## b 2.791449e-03 2.482816e-03 3.164128e-03
## clx 2.405288e+02 2.115740e+02 2.776275e+02
plotPredy(data = paper3_uas_ndre_data,
         x = N_level_kgha,
         y = SI,
         model = modelS3_3,
         xlab = "Pre-plant N Rate",
         ylab = "Sufficiency-Index")
```



xS3_3 <- residuals(modelS3_3)
plotNormalHistogram(xS3_3)</pre>





```
aS3_3 <- summary(modelS3_3)$coefficients[1]
bS3_3 <- summary(modelS3_3)$coefficients[2]
clxS3_3 <- summary(modelS3_3)$coefficients[3]
clxS3_3_SE <- summary(modelS3_3)$coefficients[6]

plateauS3_3 <- aS3_3 + bS3_3 * clxS3_3 + (-0.5*bS3_3)*clxS3_3
plateauS3_3
```

```
## [1] 0.9931276
```

```
Fig6 <- ggplot(data = figS3_data , aes ( x = N_rate , y = SI)) +
  geom_point(data = figS3_data , aes ( x = N_rate , y = SI , shape = Index , color = Index) , size = 6
  geom_line(data = figS3_data , aes( x = N_rate , y = Fit , color = Index , linetype = Index) , size = coord_cartesian(ylim = c(0 , 1) , xlim = c(0 , 252)) +
  scale_x_continuous(breaks = seq(0 , 250 , by = 50)) +</pre>
```

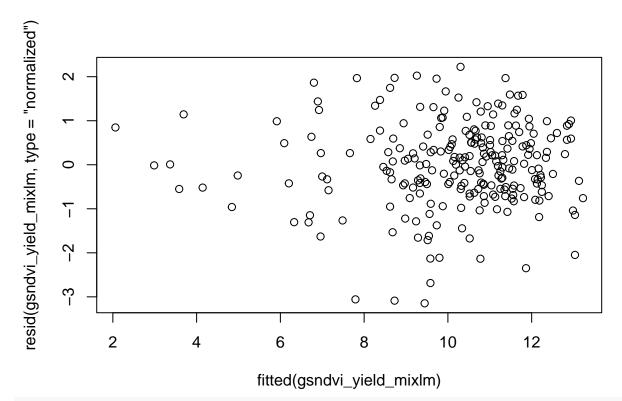
```
scale_y_continuous(breaks = seq(0, 1, by = .1)) +
    theme_classic() +
    labs(x = "Pre-plant N Rate ( kg N ha"^-1~")" , y = "Sufficiency-Index" , shape = "Index" , color = "In
    guides(color = guide_legend(keywidth = 10 , keyheight = 1.5 , unit = "cm" , title.hjust = 0.5 , byrow
                     shape = guide_legend(keywidth = 10 , keyheight = 1.5 , unit = "cm" , title.hjust = 0.5 , byrow
                     linetype = guide_legend(keywidth = 10 , keyheight = 1.5 , unit = "cm" , title.hjust = 0.5 , by
    theme(axis.title = element_text(size = 32),
                  axis.text = element_text(size = 32),
                  legend.text = element_text(size = 32),
                  legend.title = element_text(size = 32 , hjust = 0.5),
                   legend.text.align = 0,
                   legend.box.background = element_rect(size = 1),
                   legend.position = c(0.3, 0.3) +
    geom_vline(xintercept = clxS3_1 , linetype = "solid" , size = 1.5 , color = "black") +
    geom_vline(xintercept = clxS3_2 , linetype = "longdash" , size = 1.5 , color = "dodgerblue") +
    geom_vline(xintercept = clxS3_3 , linetype = "dashed" , size = 1.5 , color = "deeppink") +
    scale_shape_manual("Index", breaks = c("uas_NDVI", "gs_NDVI", "uas_NDRE"), values = c(1,2,0) scale_color_manual("Index", breaks = c("uas_NDVI", "gs_NDVI", "uas_NDRE"), values = c("dodgerble")
    scale_linetype_manual("Index" , breaks = c("uas_NDVI" , "gs_NDVI" , "uas_NDRE") , values = c("longda
Fig6
Sufficiency-Ind
                                                                        Index
                                                                                               NDVI<sub>UAS</sub>
                                                                                               NDVIGS
                                                                                               NDRE<sub>UAS</sub> 200 250
                          Pre-plant N Rate (kg N ha<sup>-1</sup>
```

gs ndvi SI

ggsave("FIGURES/Fig6.tiff" , Fig6, compression = "lzw" , width = 17 , height = 12 , type = "cairo" , dp

```
ctrl <- lmeControl(opt = "optim") #changes control to "optimal" settings</pre>
gsndvi_yield_mixlm <- lme(fixed = GrainYield_Mgha ~ SI ,</pre>
                          random = ~I(SI) | site_year,
                          data = paper3_gsdata)
summary(gsndvi_yield_mixlm)
model
## Linear mixed-effects model fit by REML
     Data: paper3_gsdata
                  BIC
##
         AIC
                          logLik
##
     644.9422 665.7506 -316.4711
##
## Random effects:
## Formula: ~I(SI) | site_year
## Structure: General positive-definite, Log-Cholesky parametrization
##
              StdDev
                         Corr
## (Intercept) 2.7563897 (Intr)
              3.2122367 -0.93
## I(SI)
## Residual
              0.8094868
##
## Fixed effects: GrainYield_Mgha ~ SI
                 Value Std.Error DF t-value p-value
## (Intercept) 2.078638 0.9363706 228 2.219888 0.0274
              9.150682 1.0833489 228 8.446662 0.0000
## Correlation:
##
      (Intr)
## SI -0.937
##
## Standardized Within-Group Residuals:
           Min
                        Q1
                                   Med
                                                QЗ
## -3.14633363 -0.54585540 0.01649065 0.59853697 2.22413109
##
## Number of Observations: 239
## Number of Groups: 10
summary(gsndvi_yield_mixlm)$tTable
##
                 Value Std.Error DF t-value
## (Intercept) 2.078638 0.9363706 228 2.219888 2.741090e-02
              9.150682 1.0833489 228 8.446662 3.538671e-15
Anova(gsndvi_yield_mixlm , type = 2)
## Analysis of Deviance Table (Type II tests)
## Response: GrainYield_Mgha
      Chisq Df Pr(>Chisq)
## SI 71.346 1 < 2.2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

```
Anova(gsndvi_yield_mixlm , type = 3)
## Analysis of Deviance Table (Type III tests)
##
## Response: GrainYield_Mgha
##
                  Chisq Df Pr(>Chisq)
                              0.02643 *
## (Intercept) 4.9279 1
## SI
               71.3461 1
                              < 2e-16 ***
##
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
r.squaredGLMM(gsndvi_yield_mixlm)
## Warning: 'r.squaredGLMM' now calculates a revised statistic. See the help page.
##
              R2m
## [1,] 0.5423214 0.8541967
gsndvi_yield_slope <- round((summary(gsndvi_yield_mixlm)$tTable[2] * 0.1) , digits = 1)</pre>
gsndvi_yield_slope
## [1] 0.9
plot (gsndvi_yield_mixlm)
                                                                 00
                   2
                                                                     0
                                                             0
                                                                                   O
                                                       O
                                    0
             Standardized residuals
                   1
                                                 0
                          0
                                                      0
                                                  0
                                                        0
                                                                                           0
                   0
                                            0
                                                   0
                                   0 0
                                                                                              0
                                           0
                  -1
                                                    08
                                                                            0
                                                        O
                                                                             0
                  -2
                                                                                            0
                                                                       00
                                                                               O
                                                                                     0
                                                                       0
                  -3
                                                            0
                                                                  0
                          2
                                                  6
                                      4
                                                              8
                                                                          10
                                                                                      12
                                                      Fitted values
diagnostics
plot(resid(gsndvi_yield_mixlm, type = "normalized") ~fitted(gsndvi_yield_mixlm))
```

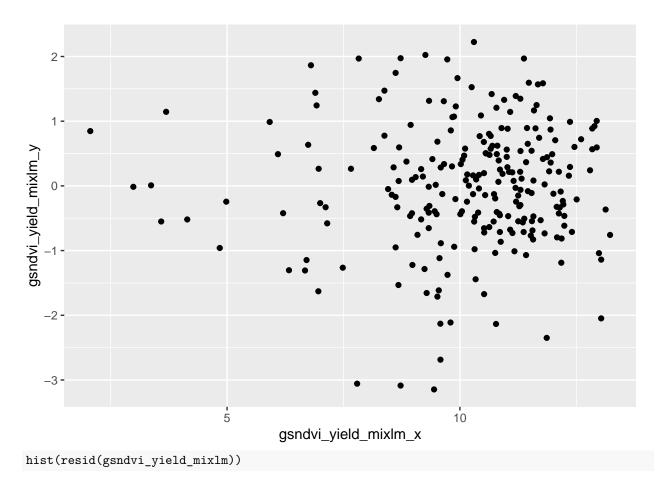


```
gsndvi_yield_mixlm_y <- resid(gsndvi_yield_mixlm, type = "normalized")
gsndvi_yield_mixlm_x <- fitted(gsndvi_yield_mixlm)

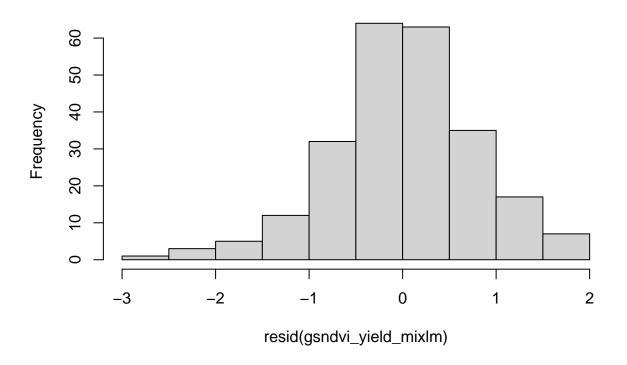
gsndvi_yield_mixlmresid_data <- data.frame(gsndvi_yield_mixlm_x , gsndvi_yield_mixlm_y)

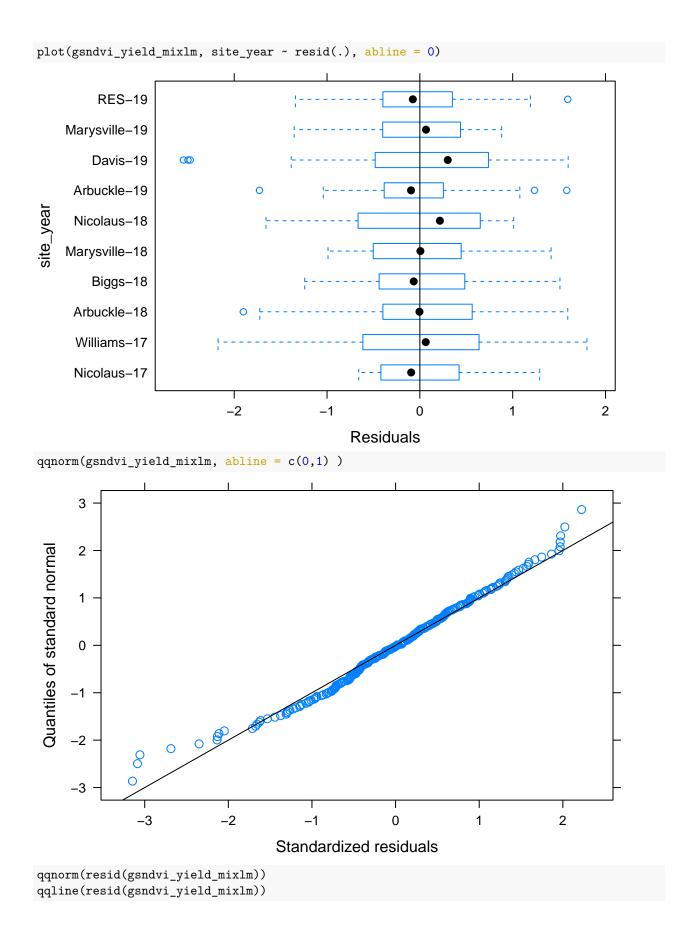
ggplot( data = gsndvi_yield_mixlmresid_data , aes( x = gsndvi_yield_mixlm_x , y = gsndvi_yield_mixlm_y)

geom_point(mapping = aes(gsndvi_yield_mixlm_x , gsndvi_yield_mixlm_y) , data = gsndvi_yield_mixlmresid_mixlm_y</pre>
```

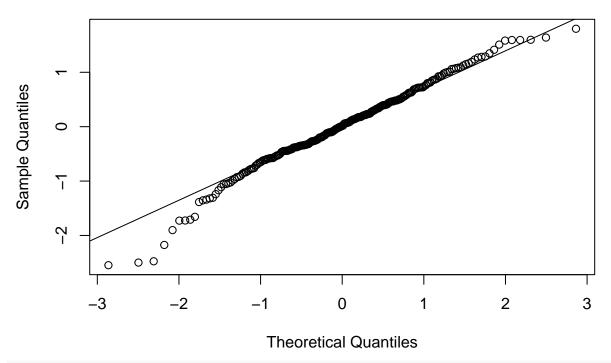


Histogram of resid(gsndvi_yield_mixlm)

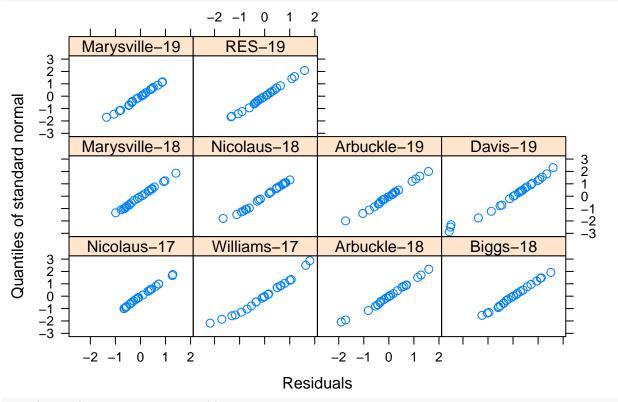




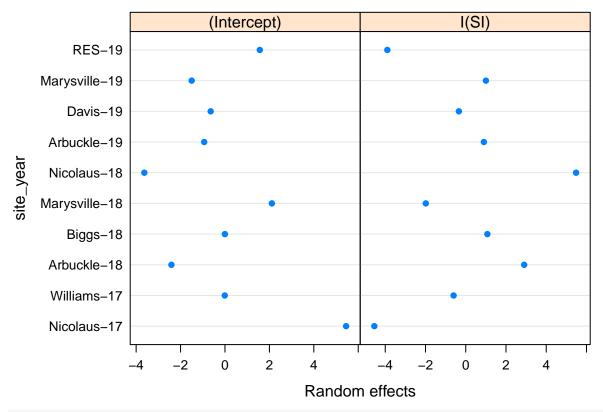
Normal Q-Q Plot

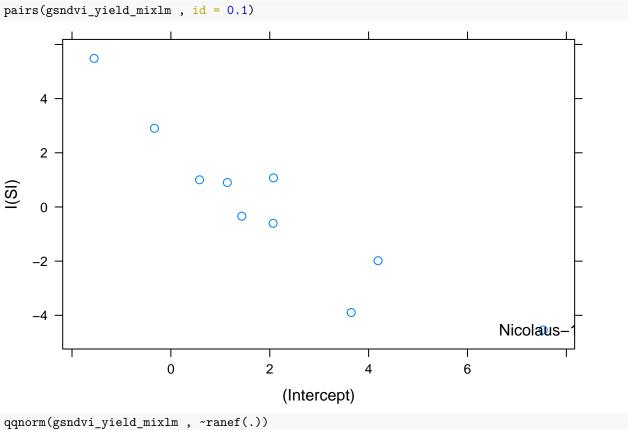


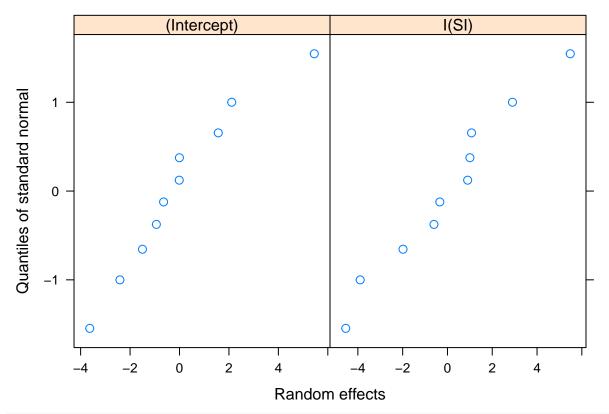
qqnorm(gsndvi_yield_mixlm , ~resid(.) | site_year)



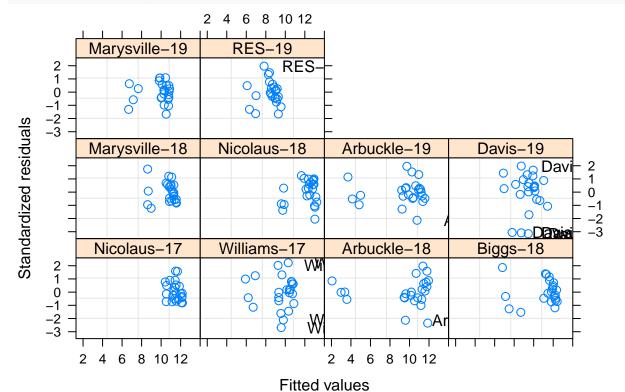
plot(ranef(gsndvi_yield_mixlm))







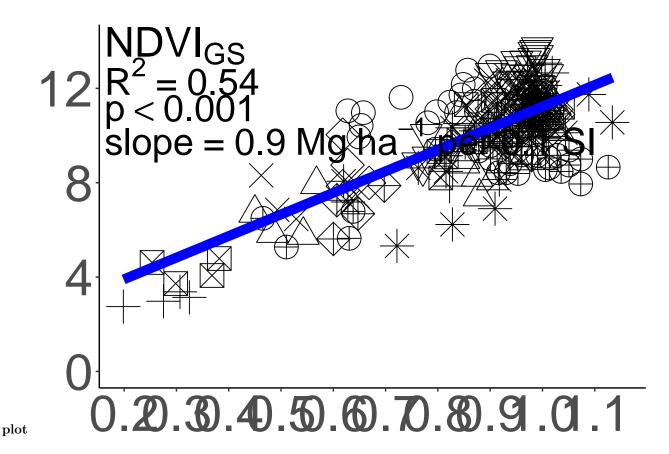
plot(gsndvi_yield_mixlm, resid(., type = "p") ~ fitted(.) | site_year,
 id = 0.05, adj = -0.3)



```
mylist <- list(SI=seq(round(min(paper3_gsdata$SI), digits = 3) , round(max(paper3_gsdata$SI), digits = 3)</pre>
gsndvi_emmeans <- as.data.frame(summary(emmeans(gsndvi_yield_mixlm , ~ SI , at = mylist )))
emmeans
plot_si1 <- ggplot( data = paper3_gsdata , aes( x = SI , y = GrainYield_Mgha)) +</pre>
     geom_point(data = paper3_gsdata, aes(x = SI, y = GrainYield_Mgha, shape = site_year), size = 8)
     geom_line( data = gsndvi_emmeans , aes(x = SI , y = emmean) , color = "blue" , size = 3.5) + 
     theme_classic() +
     labs( x = NULL , y = NULL , color = "Index") +
     scale_x_continuous(breaks = seq(.2 , 1.2 , by = .1)) +
     scale_y\_continuous(breaks = seq(0 , 14 , by = 4)) +
     coord_cartesian(ylim = c(0,14), xlim = c(.2, 1.15)) +
     theme(axis.text = element_text(size = 38),
                   axis.title = element_text(size = 38),
                   legend.text = element_text(size = 24),
                   legend.title = element_text(size = 24),
                    legend.position = "none") +
     annotate("text", x=0.1625, y=13.8, label="NDVI[GS]", size=11 , color = "black" , hjust = 0 , parse = '
     annotate("text", x=0.1625, y=12.5, label="R^2-"="~0.54", size=10 , color="black", hjust=0 , parse label="label" | label="lab
     annotate("text", x=0.1625, y=11, label="p < 0.001", size=10 , color= "black" , hjust = 0 , parse = T)
     annotate("text", x=0.1625, y=9.8, label="slope~'='~0.9~Mg~ha^-1~per~0.1~SI", size=10 , color= "black"
```

scale_shape_manual(values = c(1:20))

plot_si1

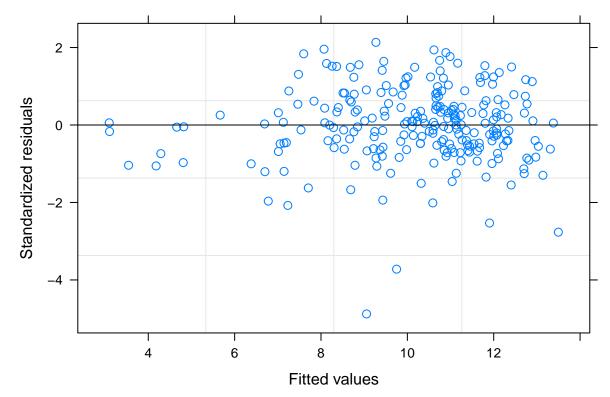


sUAS ndre RI

model

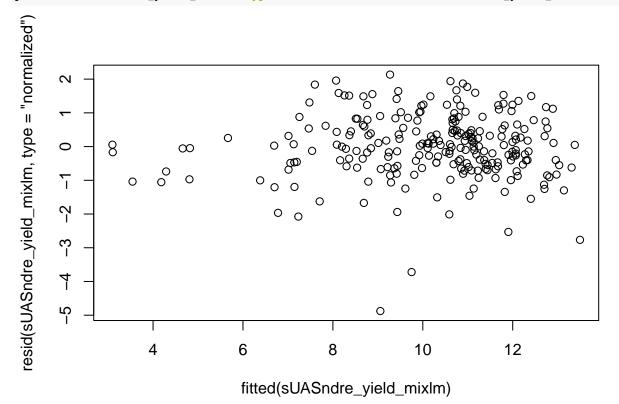
```
## Linear mixed-effects model fit by REML
    Data: paper3_uas_ndre_data
                         logLik
##
         AIC
                 BIC
##
     622.6791 643.4874 -305.3395
##
## Random effects:
## Formula: ~I(SI) | site_year
   Structure: General positive-definite, Log-Cholesky parametrization
##
              StdDev
                        Corr
## (Intercept) 3.6233290 (Intr)
              3.9879705 -0.949
## I(SI)
## Residual
              0.7664143
##
## Fixed effects: GrainYield_Mgha ~ SI
                  Value Std.Error DF t-value p-value
## (Intercept) 0.133149 1.206608 228 0.110350 0.9122
## SI
              11.320107 1.328094 228 8.523573 0.0000
```

```
## Correlation:
##
     (Intr)
## SI -0.953
##
## Standardized Within-Group Residuals:
           Min
                       Q1
                                                 QЗ
                                   Med
                                                            Max
## -4.877388867 -0.548696368 0.007240419 0.548771256 2.133897894
##
## Number of Observations: 239
## Number of Groups: 10
summary(sUASndre_yield_mixlm)$tTable
                   Value Std.Error DF t-value
                                                    p-value
## (Intercept) 0.1331486 1.206608 228 0.1103495 9.122293e-01
              Anova(sUASndre_yield_mixlm , type = 2)
## Analysis of Deviance Table (Type II tests)
##
## Response: GrainYield_Mgha
      Chisq Df Pr(>Chisq)
## SI 72.651 1 < 2.2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Anova(sUASndre_yield_mixlm , type = 3)
## Analysis of Deviance Table (Type III tests)
## Response: GrainYield_Mgha
               Chisq Df Pr(>Chisq)
## (Intercept) 0.0122 1
                            0.9121
              72.6513 1
## SI
                            <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
r.squaredGLMM(sUASndre_yield_mixlm)
             R2m
## [1,] 0.5051416 0.8667079
uas_ndre_yield_slope <- round((summary(sUASndre_yield_mixlm)$tTable[2] * 0.1) , digits = 1)</pre>
uas_ndre_yield_slope
## [1] 1.1
plot (sUASndre_yield_mixlm)
```



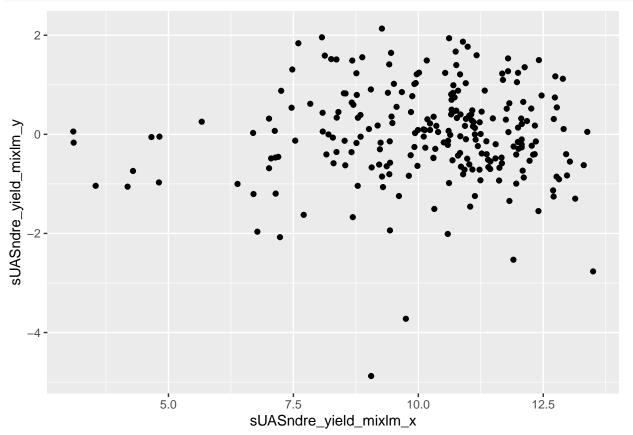
 ${\bf diagnostics}$

plot(resid(sUASndre_yield_mixlm, type = "normalized") ~fitted(sUASndre_yield_mixlm))



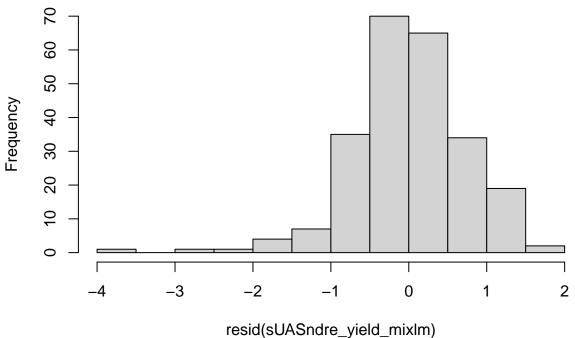
sUASndre_yield_mixlm_y <- resid(sUASndre_yield_mixlm, type = "normalized")
sUASndre_yield_mixlm_x <- fitted(sUASndre_yield_mixlm)</pre>

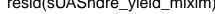
sUASndre_yield_mixlmresid_data <- data.frame(sUASndre_yield_mixlm_x , sUASndre_yield_mixlm_y)</pre>

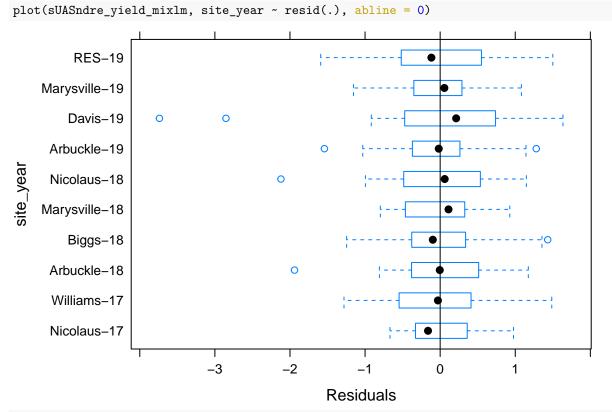


hist(resid(sUASndre_yield_mixlm))

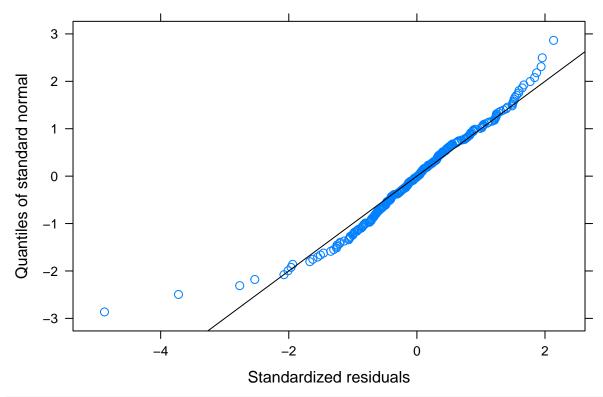
Histogram of resid(sUASndre_yield_mixlm)





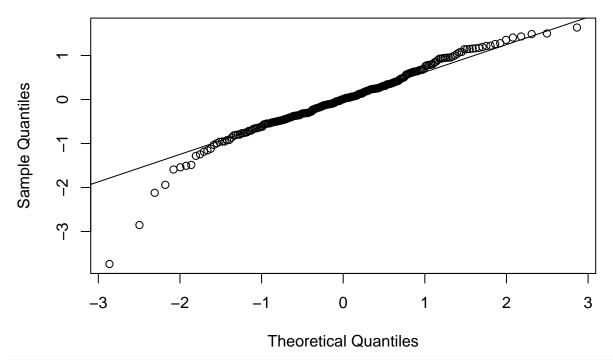


qqnorm(sUASndre_yield_mixlm, abline = c(0,1))

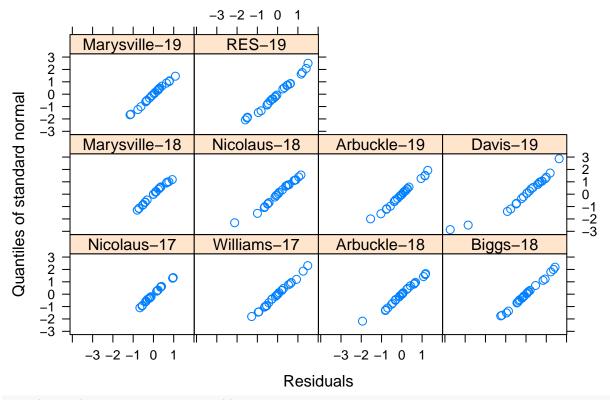


qqnorm(resid(sUASndre_yield_mixlm))
qqline(resid(sUASndre_yield_mixlm))

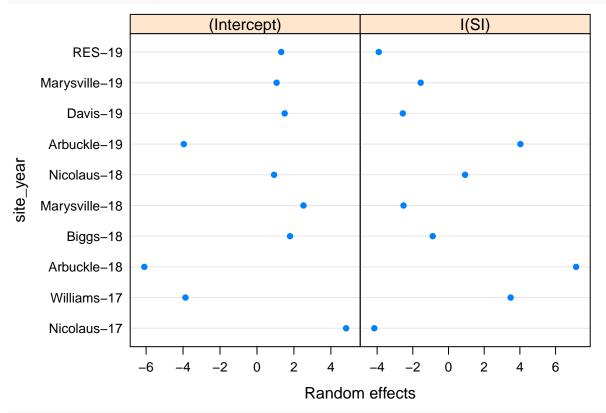
Normal Q-Q Plot



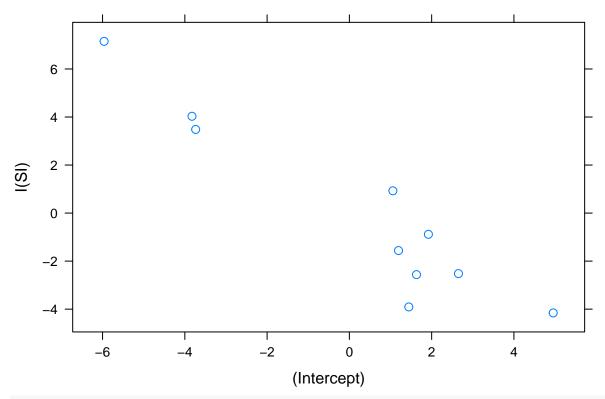
qqnorm(sUASndre_yield_mixlm , ~resid(.) | site_year)



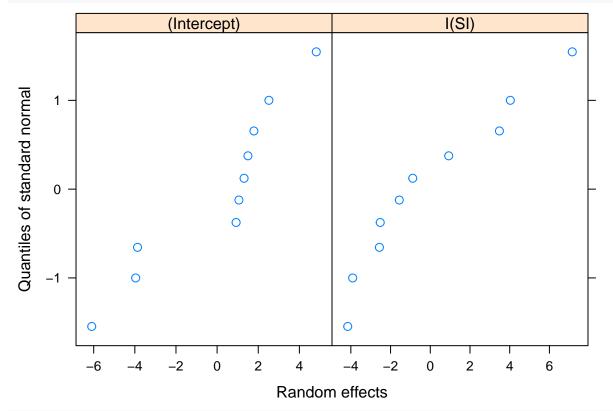
plot(ranef(sUASndre_yield_mixlm))



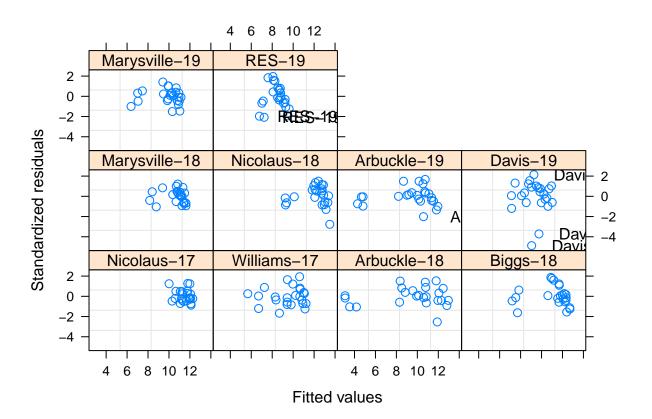
pairs(sUASndre_yield_mixlm , id = 0.1)



qqnorm(sUASndre_yield_mixlm , ~ranef(.))



plot(sUASndre_yield_mixlm, resid(., type = "p") ~ fitted(.) | site_year,
 id = 0.05, adj = -0.3)

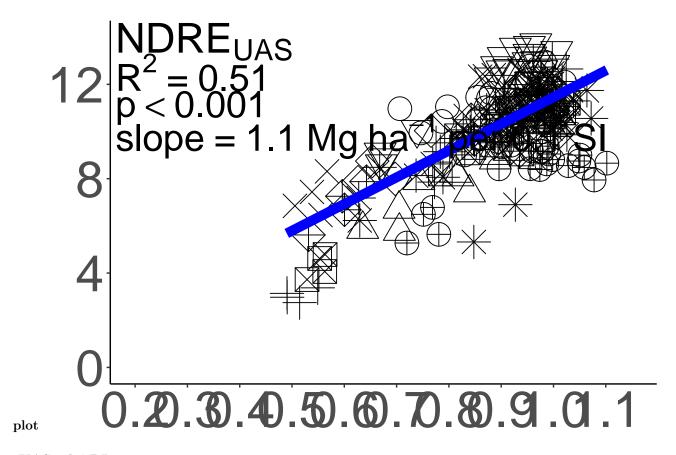


emmeans

plot_si2

```
mylist <- list(SI=seq(round(min(paper3_uas_ndre_data$SI), digits = 3) , round(max(paper3_uas_ndre_data$
sUASndre_emmeans <- as.data.frame(summary(emmeans(sUASndre_yield_mixlm , ~ SI , at = mylist )))</pre>
```

```
plot_si2 \leftarrow ggplot( data = paper3_uas_ndre_data , aes( x = SI , y = GrainYield_Mgha)) +
  geom_point( data = paper3_uas_ndre_data , aes( x = SI , y = GrainYield_Mgha , shape = site_year) , si
  geom_line( data = sUASndre_emmeans , aes(x = SI , y = emmean) , color = "blue" , size = 3.5) +
  theme_classic() +
  labs( x = NULL , y = NULL , shape = "Site-Year") +
  scale_x_continuous(breaks = seq(0.2, 1.2, by = .1)) +
  scale_y\_continuous(breaks = seq(0 , 14 , by = 4)) +
  coord_cartesian(ylim = c(0,14), xlim = c(0.2, 1.15)) +
  theme(axis.text = element_text(size = 38),
        axis.title = element_text(size = 38),
        legend.text = element_text(size = 20),
        legend.title = element_text(size = 20),
        legend.position = "none") +
  annotate("text", x=0.1625, y=13.8, label="NDRE[UAS]", size=11 , color = "black" , hjust = 0 , parse =
  annotate("text", x=0.1625, y=12.5, label="R^2~'='~0.51", size=10 , color= "black" , hjust = 0 , parse
  annotate("text", x=0.1625, y=11, label="p < 0.001", size=10 , color= "black" , hjust = 0 , parse = T)
  annotate("text", x=0.1625, y=9.8, label="slope~'='~1.1~Mg~ha^-1~per~0.1~SI", size=10 , color= "black"
  scale_shape_manual(values = c(1:20))
```

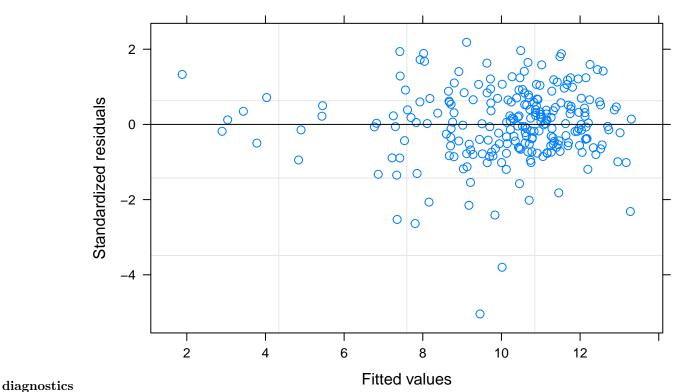


sUAS ndvi RI

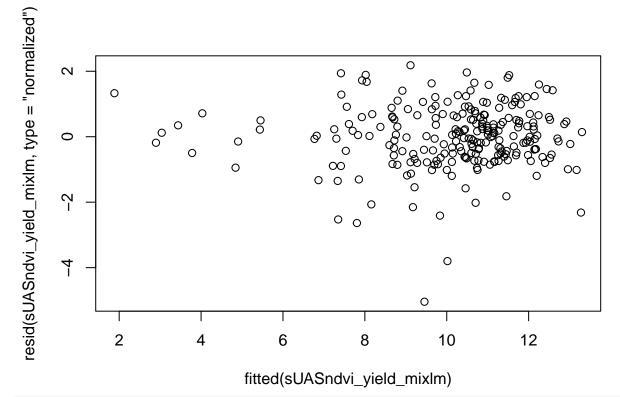
model

```
## Linear mixed-effects model fit by REML
    Data: paper3 uas ndvi data
##
        AIC
                BIC
                       logLik
##
    649.121 669.9294 -318.5605
##
## Random effects:
## Formula: ~I(SI) | site_year
## Structure: General positive-definite, Log-Cholesky parametrization
##
              StdDev
                        Corr
## (Intercept) 8.0227182 (Intr)
              8.6254055 -0.992
## I(SI)
## Residual
              0.8208589
##
## Fixed effects: GrainYield_Mgha ~ SI
                  Value Std.Error DF
                                       t-value p-value
## (Intercept) -13.24868 2.740879 228 -4.833734
## SI
              24.40018 2.930391 228 8.326593
```

```
## Correlation:
##
      (Intr)
## SI -0.993
##
## Standardized Within-Group Residuals:
                       Q1
          Min
                                 Med
                                             Q3
                                                         Max
## -5.04097534 -0.54083891 0.02064131 0.59360607 2.18271187
## Number of Observations: 239
## Number of Groups: 10
summary(sUASndvi_yield_mixlm)$tTable
                  Value Std.Error DF
                                      t-value
                                                    p-value
## (Intercept) -13.24868 2.740879 228 -4.833734 2.462586e-06
               24.40018 2.930392 228 8.326593 7.720481e-15
Anova(sUASndvi_yield_mixlm , type = 2)
## Analysis of Deviance Table (Type II tests)
##
## Response: GrainYield_Mgha
      Chisq Df Pr(>Chisq)
## SI 69.332 1 < 2.2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Anova(sUASndvi_yield_mixlm , type = 3)
## Analysis of Deviance Table (Type III tests)
## Response: GrainYield_Mgha
               Chisq Df Pr(>Chisq)
## (Intercept) 23.365 1 1.34e-06 ***
              69.332 1 < 2.2e-16 ***
## SI
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
r.squaredGLMM(sUASndvi_yield_mixlm)
            R2m
## [1,] 0.547039 0.8592628
sUASndvi_yield_slope <- round((summary(sUASndvi_yield_mixlm)$tTable[2] * 0.1) , digits = 1)
plot (sUASndvi_yield_mixlm)
```

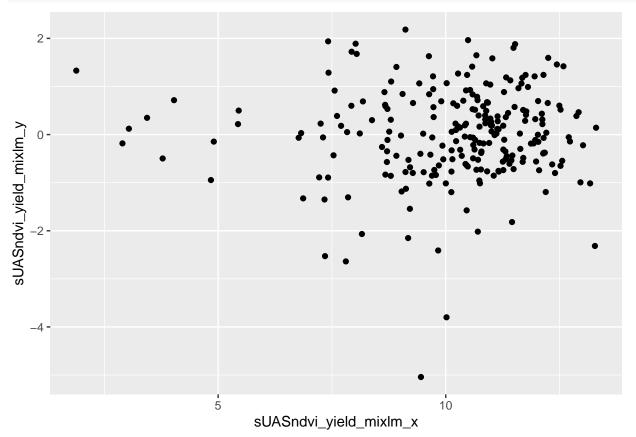


plot(resid(sUASndvi_yield_mixlm, type = "normalized") ~fitted(sUASndvi_yield_mixlm))



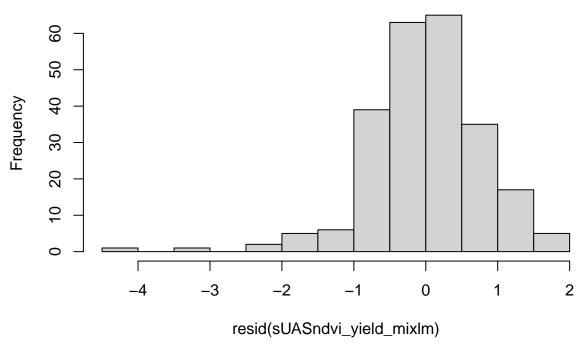
sUASndvi_yield_mixlm_y <- resid(sUASndvi_yield_mixlm, type = "normalized")
sUASndvi_yield_mixlm_x <- fitted(sUASndvi_yield_mixlm)</pre>

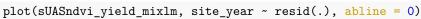
```
sUASndvi_yield_mixlmresid_data <- data.frame(sUASndvi_yield_mixlm_x , sUASndvi_yield_mixlm_y)
```

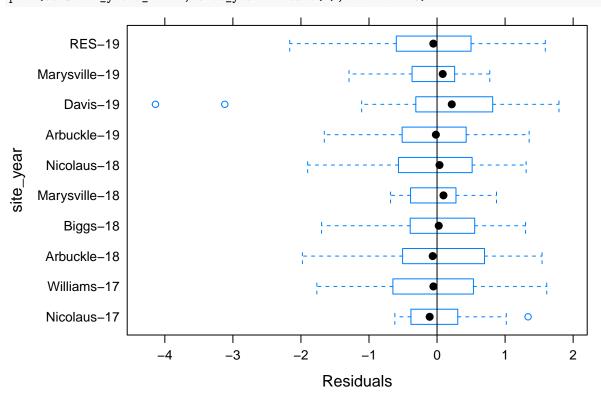


hist(resid(sUASndvi_yield_mixlm))

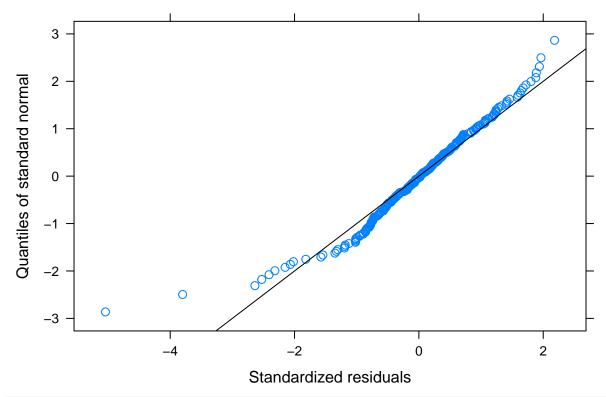
Histogram of resid(sUASndvi_yield_mixlm)





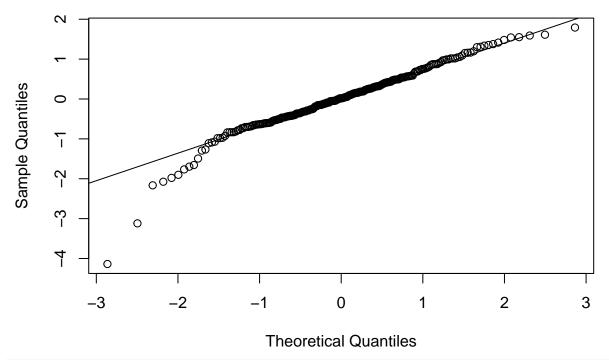


qqnorm(sUASndvi_yield_mixlm, abline = c(0,1))

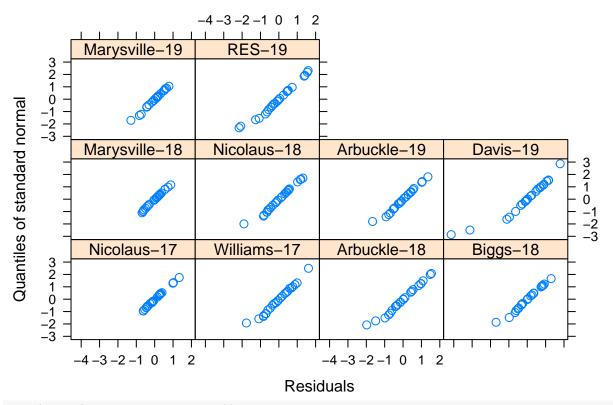


qqnorm(resid(sUASndvi_yield_mixlm))
qqline(resid(sUASndvi_yield_mixlm))

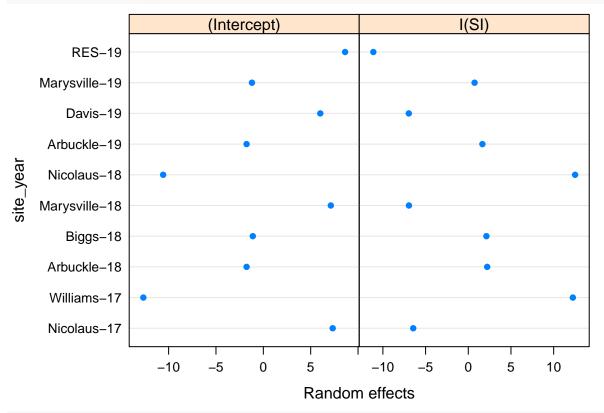
Normal Q-Q Plot



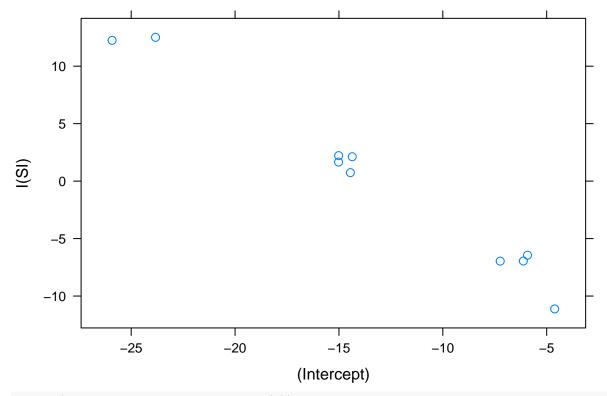
qqnorm(sUASndvi_yield_mixlm , ~resid(.) | site_year)



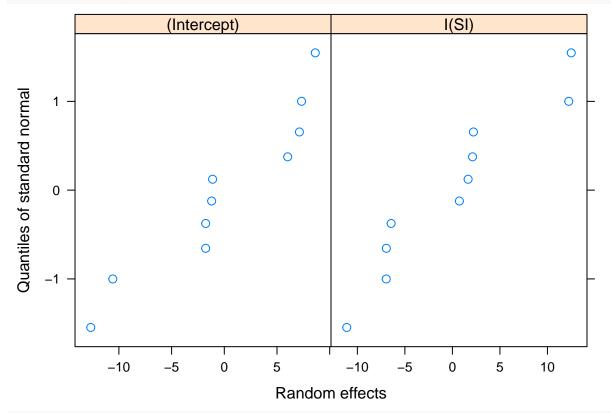
plot(ranef(sUASndvi_yield_mixlm))



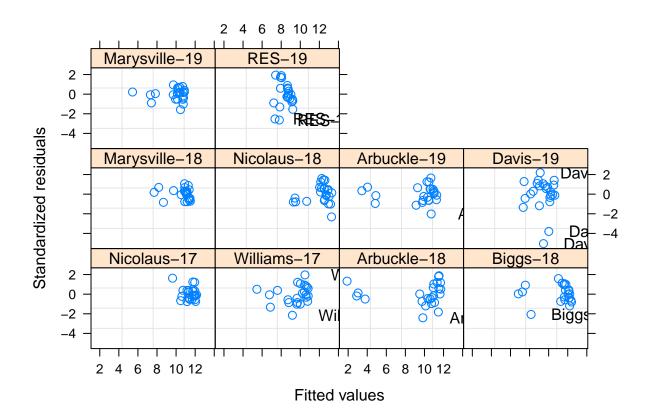
pairs(sUASndvi_yield_mixlm , id = 0.1)



qqnorm(sUASndvi_yield_mixlm , ~ranef(.))



plot(sUASndvi_yield_mixlm, resid(., type = "p") ~ fitted(.) | site_year,
 id = 0.05, adj = -0.3)



emmeans

mylist <- list(SI=seq(round(min(paper3_uas_ndvi_data\$SI), digits = 3) , round(max(paper3_uas_ndvi_data\$
sUASndvi_emmeans <- as.data.frame(summary(emmeans(sUASndvi_yield_mixlm , ~ SI , at = mylist)))</pre>

```
plot_si3 \leftarrow ggplot(data = paper3_uas_ndvi_data , aes(x = SI , y = GrainYield_Mgha)) +
  geom_point( data = paper3_uas_ndvi_data , aes( x = SI , y = GrainYield_Mgha , shape = site_year) , si
  geom_line( data = sUASndvi_emmeans , aes(x = SI , y = emmean) , color = "blue" , size = 3.5) +
  theme classic() +
  labs(x = NULL , y = NULL , shape = "Site-Year") +
  scale_x_continuous(breaks = seq(0.2, 1.2, by = .1)) +
  scale_y\_continuous(breaks = seq(0 , 14 , by = 4)) +
  coord_cartesian(ylim = c(0,14), xlim = c(0.2, 1.15)) +
  theme(axis.text = element_text(size = 38),
        axis.title = element_text(size = 38),
        legend.text = element_text(size = 20),
        legend.title = element_text(size = 20 , hjust = 0.5),
        legend.position = c(.065, .315),
        legend.box.background = element_rect(size = 1)) +
  annotate("text", x=0.1625, y=13.8, label="NDVI[UAS]", size=11 , color = "black" , hjust = 0 , parse =
  annotate("text", x=0.1625, y=12.5, label="R^2~'='~0.55", size=10 , color= "black" , hjust = 0 , parse
  annotate("text", x=0.1625, y=11, label="p < 0.001", size=10 , color= "black" , hjust = 0 , parse = T)
  annotate("text", x=0.1625, y=9.8, label="slope~'='~2.4~Mg~ha^-1~per~0.1~SI", size=10 , color= "black"
  scale_shape_manual(values = c(1:20)) +
  guides(shape = guide_legend(override.aes = list(size = 6)))
```

plot_si3

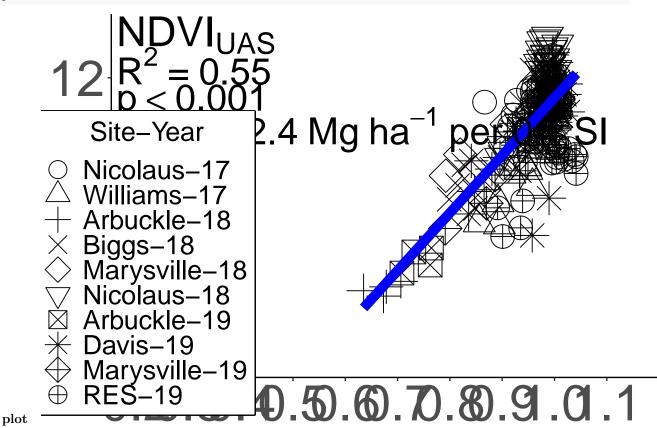
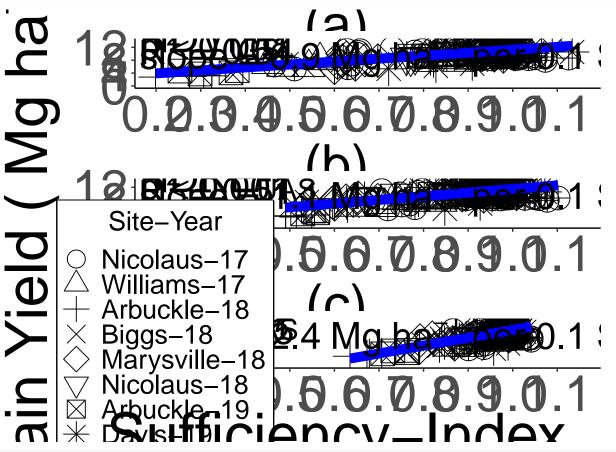


FIGURE 7



 $ggsave("FIGURES/Fig7.tiff" \ , \ Fig7 \ , \ compression = "lzw" \ , \ width = 20 \ , \ height = 22, \ type = "cairo" \ , \ dpression = "lzw" \ , \ d$