N uptake calculations

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Background

Nitrogen (N) is the currency of agronomic systems, making crop N uptake a key response variable in many research projects. The research group in the beginning of 2024 realised that N uptake calculations often did not account for moisture of samples that were at differring stages. This could lead to inaccurate computations of N upatake.

Telha Rehman performed empirical analysis, showing that samples had the following corresponding moisture contents:

Oven dry: 1.9% Ambient straw: 8.27% Ambient grain: 9.76%

The basic idea of this script - which can be adapted for everyone's individual project - is that computation involving two or more weights need to be corrected for moisture. In this script, Zhang has chosen to correct all weights to the "ambient" values.

Necessary libraries

Read and check data

```
#read data
master = read_excel('SIF_NRate_NUptake_Base_Dataframe.xlsx', sheet = 1)
master $Microbalance_sample_Weight_mg <- as.numeric(master $Microbalance_sample_Weight_mg)
str(master)
## tibble [336 x 12] (S3: tbl_df/tbl/data.frame)
                                       : chr [1:336] "701-MG" "702-MG" "703-MG" "704-MG" ...
## $ SampleID
## $ SIF_N_µg
                                       : num [1:336] 85.8 56.2 65.3 41.8 73.9 ...
## $ Microbalance_sample_Weight_mg
                                      : num [1:336] 5.5 4.94 4.98 4.72 5.23 ...
## $ Oven_dry_Sample_Weight_g
                                      : num [1:336] 5264105 9981691 7902092 8450992 5596197 ...
## $ Area_harvested_m2
                                       : num [1:336] 10000 10000 10000 10000 10000 10000 10000
## $ Oven_dry_sample_moisture_percent : num [1:336] 14 14 14 14 14 14 14 14 14 14 1...
## $ Stage
                                       : chr [1:336] "Maturity_Grain" "Maturity_Grain" "Maturity_Gr
                                       : chr [1:336] "FR" "FR" "FR" "FR" ...
## $ Treatment
## $ Blk
                                       : num [1:336] 7 7 7 7 7 7 7 7 7 7 ...
                                       : chr [1:336] "NIL" "NIL" "NIL" "NIL" ...
## $ Topdress
## $ Year
                                       : num [1:336] 2023 2023 2023 2023 ...
table(master$Stage)
##
##
         Heading Maturity_Grain Maturity_Straw
                                                      ΡI
##
             48
                          132
                                        132
                                                       24
table(master$Treatment)
##
## CR FR
## 36 36
table(master$Topdress)
##
##
    0
        1 NIL
## 96 72 168
table(master$Blk)
##
## 7 8 9
## 24 24 24
```

```
table(master$Year)

##
## 2021 2022 2023
## 132 132 72

#dataset looks good and complete
```

N uptake calculations

There were samples taken at different physiological stages for Zhang's study. If you have samples taken before maturity, please pay close attention to the choices made below. If all samples were taken at maturity, then focus on how samples at maturity can be handled.

Specifically for this data set, maturity samples were harvested (1) by hand and (2) by a combine.

For (1) harvest by hand oven dry weights were used. For (2) combine harvest, 14% moisture was used in combination with harvest index was used to calculate N uptake.

N uptake for PI and 50% heading are determined from non-segregated plant parts. i.e. the whole plant was ground to a powder together and is representative of the crop's N uptake. As most of the tissue here is vegetative, we assume that ambient moisture is the same as straw (8.27%)

N uptake for maturity are determined from separated plant parts - grain and straw. This step calculates N present in grain and straw, but do not sum them yet. Moisture for is 8.27% for straw and 9.76% for grains.

Dry samples were measured after stabilizing in the oven for a period of time (>24hrs) and was measured to be 1.9%.

For 2023, we directly used yield. Grain yield was reported at 14%

\$ Oven_dry_sample_moisture_percent

For 2023, grain was harvested by combine. A sub-sample was taken to evaluate N uptake. Harvest index (HI) was quantified from this subsample and used to estimate straw weight.

For for information on calculating weights based on moisture: https://edis.ifas.ufl.edu/publication/AG442

```
#Getting N concentration: g N per g of microbalanced sample (g/g)
master $N_per_g_microbalanced_sample <- ((master $SIF_N_µg)/(master $Microbalance_sample_Weight_mg))/1000
#Dry sample weight is at 1.9% for 2021 and 2022 samples, 14% for 2023 samples. These values were pre po
#We need to correct the weight for ambient moisture where the samples were measured into tin capsules d
\verb|master$Ambient_Sample_Weight_g| <- master | $0 ven_dry_Sample_Weight_g | $(100-master | $0 ven_dry_sample_moisture) | $(100-master | $0 ven_dry_sample_m
str(master)
## tibble [336 x 14] (S3: tbl_df/tbl/data.frame)
                                                                                                                                        : chr [1:336] "701-MG" "702-MG" "703-MG" "704-MG" ...
            $ SampleID
        $ SIF_N_μg
                                                                                                                                        : num [1:336] 85.8 56.2 65.3 41.8 73.9 ...
        $ Microbalance_sample_Weight_mg
                                                                                                                                        : num [1:336] 5.5 4.94 4.98 4.72 5.23 ...
           $ Oven_dry_Sample_Weight_g
                                                                                                                                        : num [1:336] 5264105 9981691 7902092 8450992 5596197 ...
##
                                                                                                                                        : num [1:336] 10000 10000 10000 10000 10000 10000 10000
## $ Area_harvested_m2
```

: num [1:336] 14 14 14 14 14 14 14 14 14 14 ...

```
## $ Stage
                                       : chr [1:336] "Maturity_Grain" "Maturity_Grain" "Maturity_Gr
## $ Treatment
                                       : chr [1:336] "FR" "FR" "FR" "FR" ...
## $ Blk
                                       : num [1:336] 7 7 7 7 7 7 7 7 7 7 ...
                                       : chr [1:336] "NIL" "NIL" "NIL" "NIL" ...
## $ Topdress
## $ Year
                                       : num [1:336] 2023 2023 2023 2023 2023 ...
                                       : num [1:336] 0.0156 0.01139 0.01311 0.00885 0.01412 ...
## $ N_per_g_microbalanced_sample
                                       : num [1:336] 5016767 9512693 7530806 8053916 5333255 ...
## $ Ambient_Sample_Weight_g
#Total N uptake = g N per g of microbalanced sample (g/g)*total sample weight (g)/area harvested (m2)
#then scale from g/m2 to kg/ha
master$N_total_kgha <- ((master$N_per_g_microbalanced_sample*master$Ambient_Sample_Weight_g)/1000) * (1
str(master)
## tibble [336 x 15] (S3: tbl_df/tbl/data.frame)
## $ SampleID
                                       : chr [1:336] "701-MG" "702-MG" "703-MG" "704-MG" ...
## $ SIF_N_µg
                                       : num [1:336] 85.8 56.2 65.3 41.8 73.9 ...
## $ Microbalance_sample_Weight_mg
                                       : num [1:336] 5.5 4.94 4.98 4.72 5.23 ...
## $ Oven_dry_Sample_Weight_g
                                       : num [1:336] 5264105 9981691 7902092 8450992 5596197 ...
                                       : num [1:336] 10000 10000 10000 10000 10000 10000 10000
## $ Area_harvested_m2
                                     : num [1:336] 14 14 14 14 14 14 14 14 14 ...
## $ Oven_dry_sample_moisture_percent
## $ Stage
                                       : chr [1:336] "Maturity_Grain" "Maturity_Grain" "Maturity_Gr
                                        : chr [1:336] "FR" "FR" "FR" "FR" ...
## $ Treatment
                                       : num [1:336] 7 7 7 7 7 7 7 7 7 7 ...
## $ Blk
                                       : chr [1:336] "NIL" "NIL" "NIL" "NIL" ...
## $ Topdress
## $ Year
                                       : num [1:336] 2023 2023 2023 2023 ...
## $ N_per_g_microbalanced_sample
                                       : num [1:336] 0.0156 0.01139 0.01311 0.00885 0.01412 ...
## $ Ambient_Sample_Weight_g
                                       : num [1:336] 5016767 9512693 7530806 8053916 5333255 ...
                                       : num [1:336] 78.3 108.4 98.7 71.3 75.3 ...
## $ N_total_kgha
```

Getting the maturity summed values: summing N uptake values of straw and grain

```
#separate into straw and grains for summing
maturity_straw <- master %>% filter(Stage == ("Maturity_Straw"))
maturity_grain <- master %>% filter(Stage == ("Maturity_Grain"))

#maturity_straw <-
# maturity_straw %>%
#filter(Year != 2021) %>%
# mutate(Plot = substr(SampleID, 1, 3))

#maturity_grain <-
# maturity_straw %>%
#filter(Year != 2021) %>%
#mutate(Plot = substr(SampleID, 1, 3))

as.numeric(maturity_straw$Plot) - as.numeric(maturity_grain$Plot)
```

```
## Warning: Unknown or uninitialised column: 'Plot'.
## Unknown or uninitialised column: 'Plot'.
## numeric(0)
#Create dataframe for maturity total N uptake
maturity_summed <- as.data.frame(rbind(master[1:36,], master[109:156,], master[241:288,]))
# just so we can keep the other columns
maturity_summed$Stage <- "Maturity"</pre>
#sum grain and straw into one maturity total N uptake dataframe
maturity_summed$N_total_kgha <- maturity_straw$N_total_kgha+maturity_grain$N_total_kgha
str(maturity_summed)
                 132 obs. of 15 variables:
## 'data.frame':
## $ SampleID
                                      : chr "701-MG" "702-MG" "703-MG" "704-MG" ...
## $ SIF_N_µg
                                      : num 85.8 56.2 65.3 41.8 73.9 ...
## $ Microbalance_sample_Weight_mg
                                      : num 5.5 4.94 4.98 4.72 5.23 ...
## $ Oven_dry_Sample_Weight_g
                                      : num 5264105 9981691 7902092 8450992 5596197 ...
## $ Area_harvested_m2
                                      : num 10000 10000 10000 10000 10000 10000 10000 10000
## $ Oven dry sample moisture percent
                                   : num 14 14 14 14 14 14 14 14 14 14 ...
: chr "Maturity" "Maturity" "Maturity" "Maturity" ...
## $ Treatment
                                      : chr "FR" "FR" "FR" "FR" ...
## $ Blk
                                      : num 7777777777...
## $ Topdress
                                      : chr "NIL" "NIL" "NIL" "NIL" ...
## $ Year
                                      : num 2023 2023 2023 2023 ...
## $ N_per_g_microbalanced_sample
                                      : num
                                             0.0156 0.01139 0.01311 0.00885 0.01412 ...
## $ Ambient_Sample_Weight_g
                                      : num
                                             5016767 9512693 7530806 8053916 5333255 ...
                                      : num 163.4 165.2 175.4 95.6 151.9 ...
## $ N_total_kgha
```

Bind everything back again - this is such a pain in the rear end. Add in plots and Nrates

```
"301", "302", "303", "304", "305", "306") ~ "FR",
  Plot %in% c("107", "108", "109", "110", "111", "112",
              "207", "208", "209", "210", "211", "212",
              "307", "308", "309", "310", "311", "312") ~ "CR",
  TRUE ~ "Other" # This line handles cases where plot is not listed
  ))
master maturity summed others <-
  master_maturity_summed %>%
  filter(Year != 2021) %>%
  mutate(Plot = substr(SampleID, 1, 3)) %>%
  mutate(Treatment = case_when(
  Plot %in% c("401", "402", "403", "404", "405", "406",
              "501", "502", "503", "504", "505", "506",
              "601", "602", "603", "604", "605", "606",
              "701", "702", "703", "704", "705", "706",
              "801", "802", "803", "804", "805", "806",
              "901", "902", "903", "904", "905", "906") ~ "FR",
  Plot %in% c("407", "408", "409", "410", "411", "412",
              "507", "508", "509", "510", "511", "512",
              "607", "608", "609", "610", "611", "612",
              "707", "708", "709", "710", "711", "712",
              "807", "808", "809", "810", "811", "812",
              "907", "908", "909", "910", "911", "912") ~ "CR",
  TRUE ~ "Other" # This line handles cases where plot is not listed
  ))
master_maturity_summed_clean <- rbind(master_maturity_summed_2021, master_maturity_summed_others)
# add in blk, days
master_maturity_summed_clean <- master_maturity_summed_clean %>%
  mutate(Blk = substr(Plot, 1, 1))%>%
  mutate(Days = case_when(
  Stage %in% c("PI") ~ "47",
  Stage %in% c("Heading") ~ "82",
  Stage %in% c("Maturity") ~ "134",
  TRUE ~ "Other" # This line handles cases where plot is not listed
  ))
# read in Nrate reference file
#clean the data frame to have one plot match with one Nrate
Nrate ref <- read excel("D:/Academics/UC Davis/School Work/Linquist Lab/Data/R stats/Agronomic paper/Yi
# put in the Nrate for all the plots
master_maturity_summed_clean <- master_maturity_summed_clean %>%
 left_join(Nrate_ref %>% select(Plot, Nrate_kgha), by = "Plot")
```

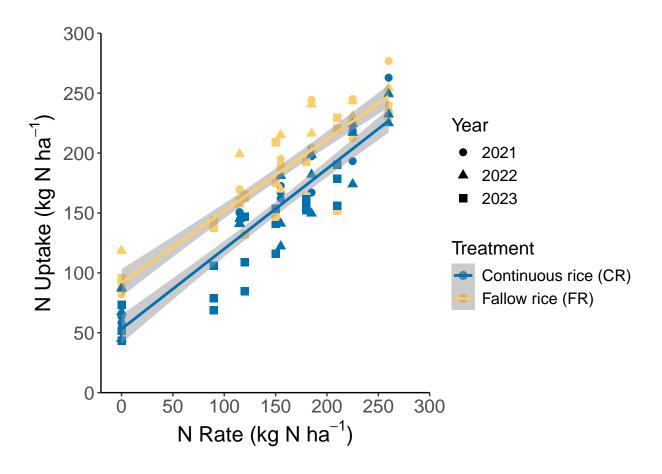
Export dataframe

```
for_export <- master_maturity_summed_clean %>%
                         select(SampleID,
                                Plot,
                                 Stage,
                                 Treatment,
                                Year,
                                 Topdress,
                                 N_total_kgha,
                                 Blk,
                                 Days,
                                 Nrate_kgha)
head(master_maturity_summed_clean)
## # A tibble: 6 x 18
##
     SampleID SIF_N_µg Microbalance_sample_Weight_mg Oven_dry_Sample_Weight_g
                 <dbl>
                                                <dbl>
## 1 RES 104
                  79.3
                                                 5.39
                                                                           286.
## 2 RES 201
                  99.0
                                                 5.28
                                                                           270.
                  74.4
                                                 5.25
                                                                           341.
## 3 RES 304
## 4 RES 112
                  79.6
                                                 5.50
                                                                           231.
## 5 RES 212
                  73.3
                                                 5.02
                                                                           243.
## 6 RES 310
                  61.9
                                                                           235.
## # i 14 more variables: Area_harvested_m2 <dbl>,
       Oven_dry_sample_moisture_percent <dbl>,
       Microbalance ambient moisture percent <dbl>, Stage <chr>, Treatment <chr>,
## #
       Blk <chr>, Topdress <chr>, Year <dbl>, N_per_g_microbalanced_sample <dbl>,
## #
       Ambient_Sample_Weight_g <dbl>, N_total_kgha <dbl>, Plot <chr>, Days <chr>,
## #
       Nrate_kgha <dbl>
write_xlsx(for_export, "N_uptake_Maturity_Summed_Clean.xlsx")
```

Filtering for N uptake over time

Plot a small visualization to make sure things generally look good

```
Nrate_trial <- master_maturity_summed_clean %>%
  #filter(Year != 2023) %>%
  filter(Topdress!=1) %>%
  filter(Stage=="Maturity") %>%
  mutate(Year= as.factor(Year))
  #group_by(Treatment, Nrate_kgha, Year) %>%
  #mutate(N_total_kqha_se = sd(N_total_kqha)/sqrt(3)) %>%
  #summarise(N_total_kgha) = mean(N_total_kgha),
  #N_total_kgha_se = mean(N_total_kgha_se))
all_Nuptake_graph <-
ggplot(Nrate_trial, aes(x=Nrate_kgha, y=N_total_kgha, color=Treatment,shape = Year))+
  geom_point(size=2.5)+
  geom_smooth(aes(group = Treatment),method = lm, alpha=0.5)+
  scale_color_manual(values=c("#0072B2","#FFCC66"), name = "Treatment", labels = c("Continuous rice (CR
  scale_x_continuous(name=expression("N Rate (kg N ha"^{-1}*")"), limits = c(-20, 300), expand = c(0, 0
  scale_y_continuous(name=expression("N Uptake (kg N ha"^{-1}*")"), limits = c(0, 300), expand = c(0, 0
  \#geom\_errorbar(aes(ymin=N\_total\_kgha-N\_total\_kgha\_se,\ ymax=N\_total\_kgha+N\_total\_kgha\_se),\ width=3,posequents
  #geom_vline(xintercept = c(41, 50, 78, 84, 121, 136), linetype = "dashed", color = "black") +
  theme_classic()+
  theme(axis.text = element_text(size = 14), axis.title = element_text(size=16))+
  theme(legend.text = element_text(size = 12),legend.title = element_text(size = 14))+
  theme(plot.title = element_text(hjust = 0.5, size = 15))
ggsave(filename = "all_Nuptake_graph.png", # Include the file extension here
       plot = all_Nuptake_graph,
                                            # Specify the plot
       path = "C:/Users/zhang/Documents/GitHub/Rice-Agroecology-Lab/N_Uptake_Calculations",
       dpi = 400,
       height = 30, width = 30, units = "cm")
## 'geom_smooth()' using formula = 'y ~ x'
## Warning: The following aesthetics were dropped during statistical transformation: shape.
## i This can happen when ggplot fails to infer the correct grouping structure in
   the data.
## i Did you forget to specify a 'group' aesthetic or to convert a numerical
   variable into a factor?
all_Nuptake_graph
## 'geom_smooth()' using formula = 'y ~ x'
## Warning: The following aesthetics were dropped during statistical transformation: shape.
## i This can happen when ggplot fails to infer the correct grouping structure in
## i Did you forget to specify a 'group' aesthetic or to convert a numerical
   variable into a factor?
```



*plots seem to make sense. Lets proceed with more fun things!!!!