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3. Outlier detection for series of observations

4. With the cleaned data, re-do the spatial correction

# M3P Data Analysis Timepoints

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Set the right working directory.

```
rm(list = ls())
setwd("C:/Users/elise/Documents/Mémoire/Main/Data/Templates/M3P")
platform <- "M3P"
```

## Data importation

Reimport the data sets extracted from the Data Preparation and Data Analysis R Markdown.

```
list.files()
```

```
## [1] "endpoint.txt"
## [2] "plant_info.txt"
## [3] "S_timeseries.txt"
## [4] "T_timeseries.txt"
## [5] "timeseries.txt"
## [6] "timeseries_Leaf_number.txt"
## [7] "timeseries_Ligulated_leaf_number.txt"
## [8] "timeseries_Plant_emergence.txt"
## [9] "timeseries_Plant_transpiration.txt"
## [10] "timeseries_Soil_water_potential.txt"
## [11] "timeseries_Water.txt"
```

```

plant_info <- read.table("plant_info.txt", header = TRUE, sep = "\t")
endpoint <- read.table("endpoint.txt", header = TRUE, sep = "\t")
S_timeseries <- read.table("S_timeseries.txt", header = TRUE, sep = "\t")

# plant_info
plant_info <- lapply(plant_info, factor)

# endpoint
matching_cols <- intersect(names(endpoint), names(plant_info))
endpoint[, matching_cols] <- lapply(endpoint[, matching_cols], factor)
endpoint$Date <- date(endpoint$Date)
endpoint$Timestamp <- NA

# timeseries
# No data for M3P

# S_timeseries
matching_cols <- intersect(names(S_timeseries), names(plant_info))
S_timeseries[, matching_cols] <- lapply(S_timeseries[, matching_cols], factor)
S_timeseries$Timestamp <- as.POSIXct(S_timeseries$Timestamp, format = "%Y-%m-%d %H:%M:%S")
S_timeseries$date <- date(S_timeseries$date)

# T_timeseries
# No data

platform <- "M3P"

# endpoint
df <- endpoint[, colSums(is.na(endpoint)) < nrow(endpoint)]
genotype_index <- which(colnames(df) == "Genotype")
variables <- "DW_plant_g"

# timeseries
# too much data

# S_timeseries
df_S_timeseries <- S_timeseries[, colSums(is.na(S_timeseries)) < nrow(S_timeseries)]
genotype_index <- which(colnames(df_S_timeseries) == "Genotype")
variables_S <- colnames(df_S_timeseries[, c(5:(genotype_index - 1))]) # We remove the three first columns that are "Unit.ID", "Time" and "Date"

# T_timeseries
# no data

print(paste(platform, ": The variables for endpoint are", paste(variables, collapse = ", "), sep = " "))

```

```
## [1] "M3P : The variables for endpoint are DW_plant_g"
```

```
print(paste(platform, ": The variables for S_timeseries are", paste(variables_S, collapse = ", "), sep = " "))
```

```
## [1] "M3P : The variables for S_timeseries are S_Height_cm, S_Leaf_area_cmsquared"
```

```
endpoint$Plant_type <- substr(endpoint$Genotype, nchar(as.character(endpoint$Genotype)), nchar(as.character(endpoint$Genotype)))
S_timeseries$Plant_type <- substr(S_timeseries$Genotype, nchar(as.character(S_timeseries$Genotype)), nchar(as.character(S_timeseries$Genotype)))
```

Get the cleaned endpoint data

```
endpoint_clean <- endpoint
# Run the function on the dataset for all the variables
endpoint_clean <- detect_replace_outliers_by_genotype(endpoint_clean)

# We add a Plant_type variable that is H or L, with T being L
S_timeseries$Plant_type <- substr(S_timeseries$Genotype, nchar(as.character(S_timeseries$Genotype)), nchar(as.character(S_timeseries$Genotype)))

S_timeseries$Plant_type <- ifelse(S_timeseries$Plant_type %in% c("T", "L"), "Line",
                                   ifelse(S_timeseries$Plant_type == "H", "Hybrid", S_timeseries$Plant_type))
```

## Time point objects

Generation of the timePoints objects using the function “createTimePoints”.

```

timePoint_endpoint <- createTimePoints(dat = endpoint,
                                         experimentName = "EPPN2020_M3P",
                                         genotype = "Genotype",
                                         timePoint = "Date",
                                         plotId = "Unit.ID",
                                         rowNum = "Row",
                                         colNum = "Column",
                                         repId = "Replication")

timePoint_endpoint_clean <- createTimePoints(dat = endpoint_clean,
                                              experimentName = "EPPN2020_M3P",
                                              genotype = "Genotype",
                                              timePoint = "Date",
                                              plotId = "Unit.ID",
                                              rowNum = "Row",
                                              colNum = "Column",
                                              repId = "Replication")

S_timeseries <- S_timeseries %>%
  group_by(Date, Unit.ID) %>%
  slice(1) %>%
  ungroup()

timePoint_S <- createTimePoints(dat = S_timeseries,
                                 experimentName = "EPPN2020_M3P",
                                 genotype = "Genotype",
                                 timePoint = "Date",
                                 plotId = "Unit.ID",
                                 rowNum = "Row",
                                 colNum = "Column",
                                 repId = "Replication",
                                 addCheck = TRUE,
                                 checkGenotypes = "EPPN20_T")

```

```

## Warning: The following plotIds have observations for less than 50% of the time point
s:
## 148, 426

```

## Gentoypic layout

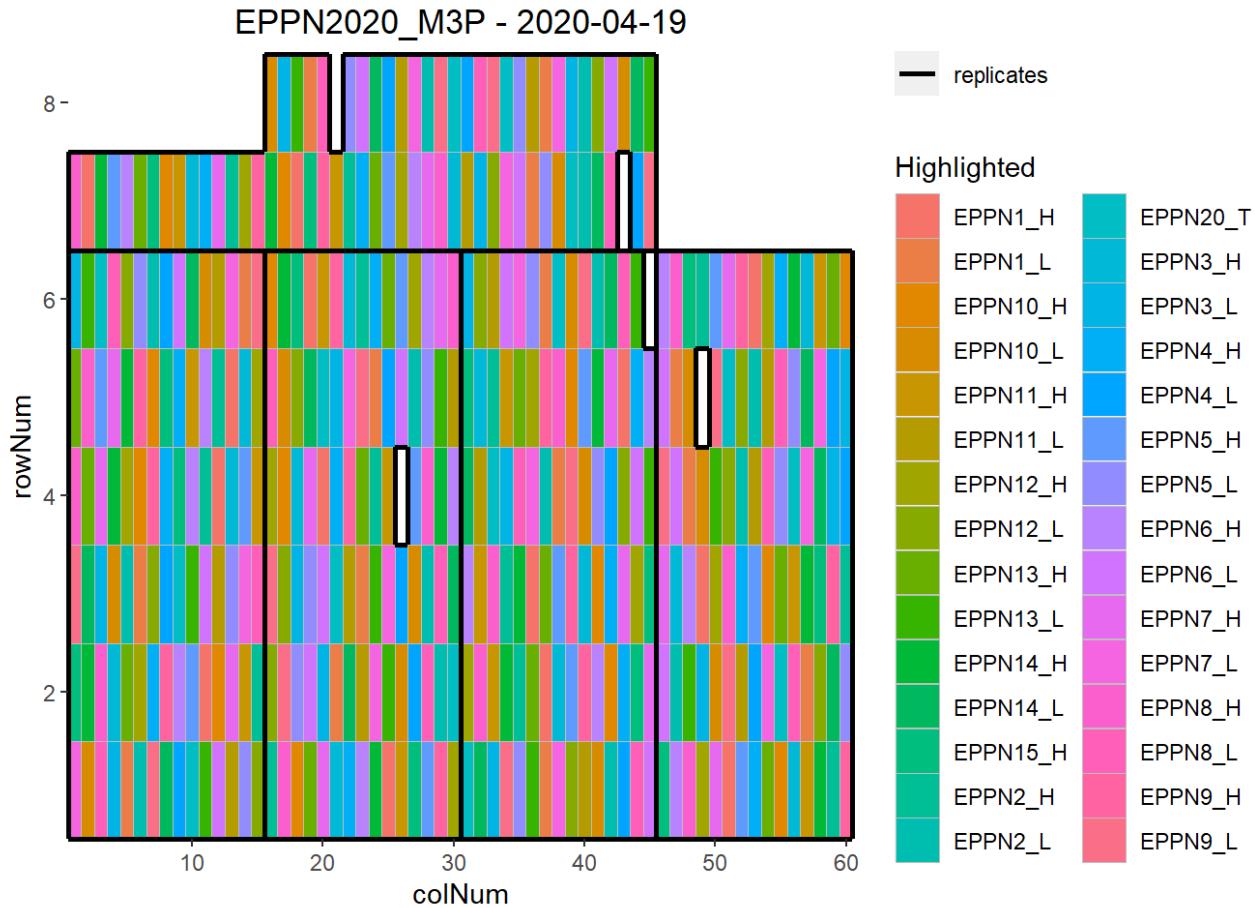
Check the layout of the platforms' genotypes.

```

genotypes_list <- as.character(unique(endpoint$Genotype))

plot(timePoint_endpoint,
      plotType = "layout",
      highlight = genotypes_list,
      showGeno = FALSE)

```



# 1. endpoint

## Comparisons between raw and cleaned data

View timePoint object.

```
summary(timePoint_endpoint)
```

```
## timePoint_endpoint contains data for experiment EPPN2020_M3P.
##
## It contains 1 time points.
## First time point: 2020-04-19
## Last time point: 2020-04-19
##
## No check genotypes are defined.
```

```
getTimePoints(timePoint_endpoint)
```

```
##   timeNumber  timePoint
## 1           1 2020-04-19
```

Count the number of observations per trait.

```
traits <- variables

for (trait_name in traits) {
  print(paste("How many data observations for", trait_name))
  num_observations <- countValid(timePoint_endpoint, trait_name)
  print(num_observations)
}
```

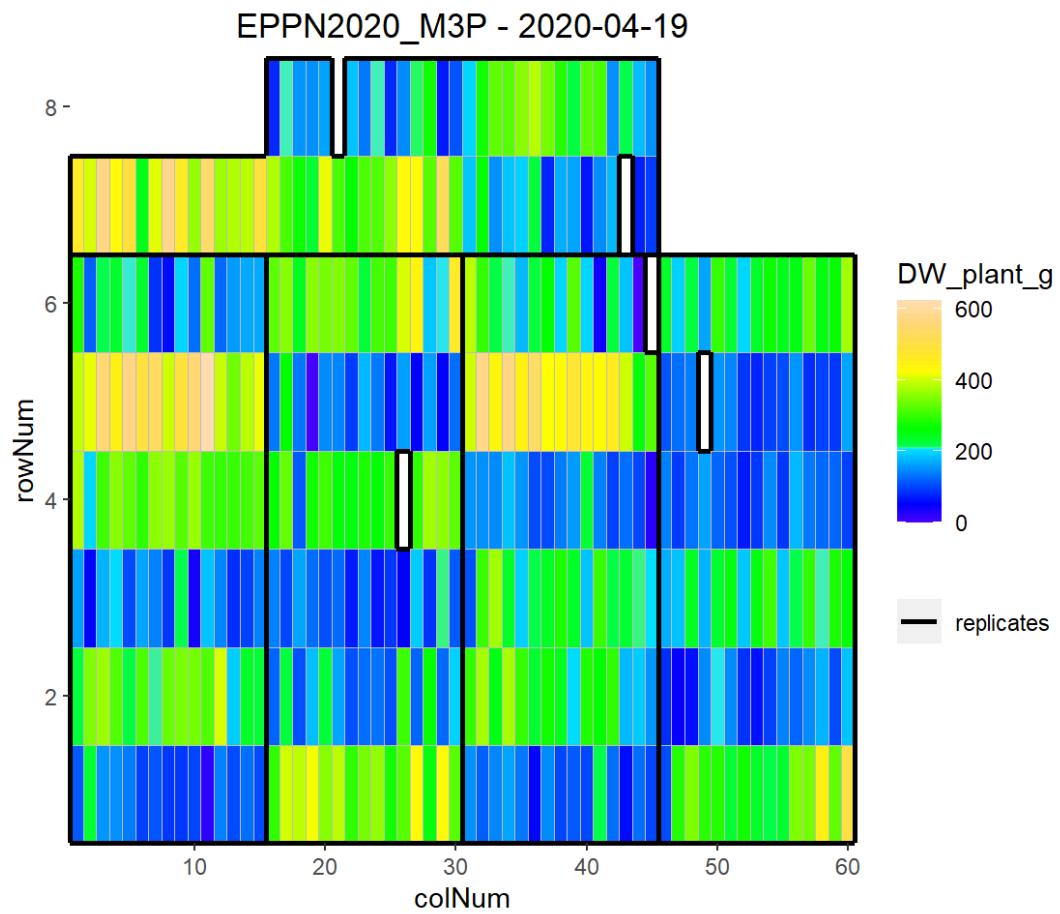
```
## [1] "How many data observations for DW_plant_g"
## 2020-04-19
##       430
```

```
for (trait_name in traits) {
  print(paste("How many cleaned data observations for", trait_name))
  num_observations <- countValid(timePoint_endpoint_clean, trait_name)
  print(num_observations)
}
```

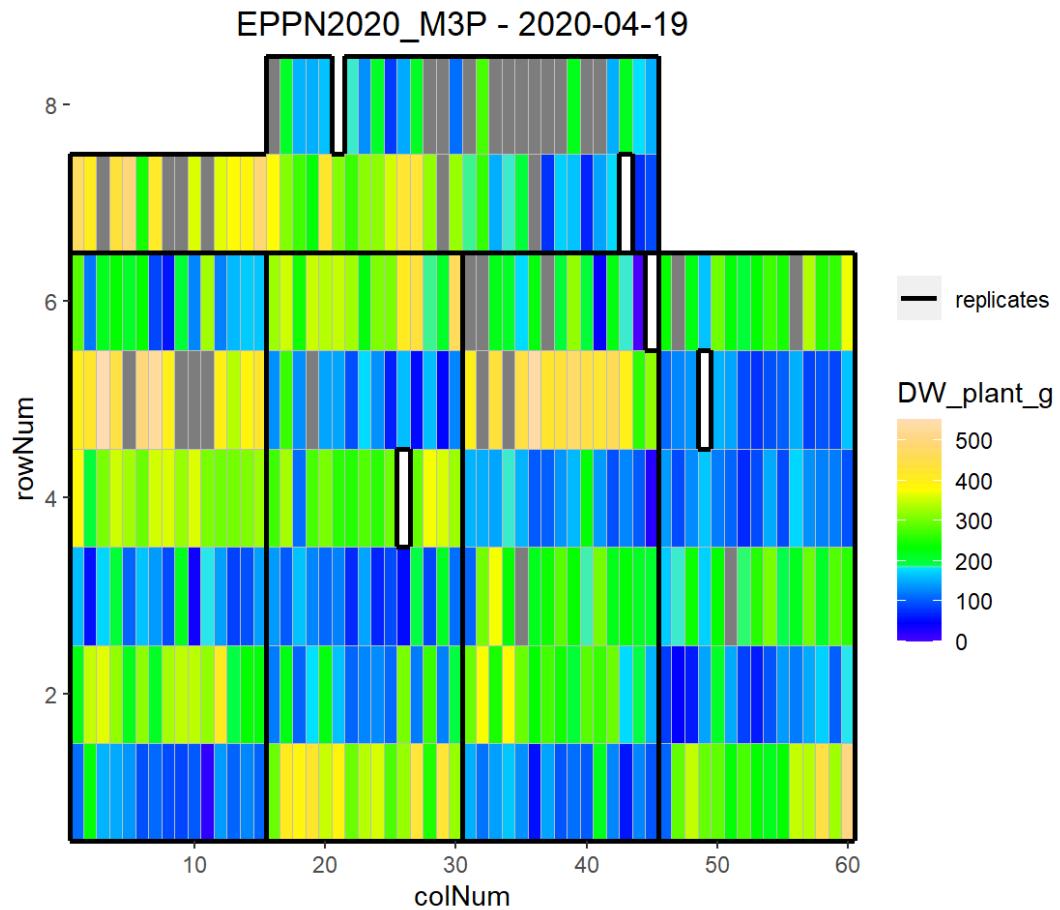
```
## [1] "How many cleaned data observations for DW_plant_g"
## 2020-04-19
##       398
```

## Check the heatmap of the data at harvest

```
for (trait_name in traits) {
  plot(timePoint_endpoint,
    plotType = "layout",
    timePoints = 1,
    traits = trait_name)
}
```



```
for (trait_name in traits) {  
  plot(timePoint_endpoint_clean,  
    plotType = "layout",  
    timePoints = 1,  
    traits = trait_name)  
}
```



## 2. S\_timeseries

### Raw data

View timePoint object

```
summary(timePoint_S)
```

```
## timePoint_S contains data for experiment EPPN2020_M3P.
##
## It contains 42 time points.
## First time point: 2020-02-14
## Last time point: 2020-04-01
##
## The following genotypes are defined as check genotypes: EPPN20_T.
```

```
getTimePoints(timePoint_S)
```

```
##   timeNumber  timePoint
## 1          1 2020-02-14
## 2          2 2020-02-15
## 3          3 2020-02-16
## 4          4 2020-02-17
## 5          5 2020-02-18
## 6          6 2020-02-19
## 7          7 2020-02-20
## 8          8 2020-02-21
## 9          9 2020-02-22
## 10        10 2020-02-23
## 11        11 2020-02-24
## 12        12 2020-02-25
## 13        13 2020-02-26
## 14        14 2020-02-28
## 15        15 2020-02-29
## 16        16 2020-03-01
## 17        17 2020-03-02
## 18        18 2020-03-03
## 19        19 2020-03-04
## 20        20 2020-03-05
## 21        21 2020-03-06
## 22        22 2020-03-07
## 23        23 2020-03-08
## 24        24 2020-03-11
## 25        25 2020-03-12
## 26        26 2020-03-13
## 27        27 2020-03-14
## 28        28 2020-03-15
## 29        29 2020-03-18
## 30        30 2020-03-19
## 31        31 2020-03-20
## 32        32 2020-03-21
## 33        33 2020-03-22
## 34        34 2020-03-24
## 35        35 2020-03-25
## 36        36 2020-03-26
## 37        37 2020-03-27
## 38        38 2020-03-28
## 39        39 2020-03-29
## 40        40 2020-03-30
## 41        41 2020-03-31
## 42        42 2020-04-01
```

```
num_timepoints <- getTimePoints(timePoint_S)
```

## Count the number of observations per trait and time point

We focus on the Height [cm] and Leaf area, because these are the two most common among the platforms.

Height is computed for 6 platforms out of 9 and area for 4 out of 9.

```

var_voulues <- c(variables_S[1], variables_S[2])
traits <- var_voulues

for (trait_name in traits) {
  print(paste("How many observations for", trait_name))
  valid_count <- countValid(timePoint_S, trait_name)
  print(valid_count)
}

```

```

## [1] "How many observations for S_Height_cm"
## 2020-02-14 2020-02-15 2020-02-16 2020-02-17 2020-02-18 2020-02-19 2020-02-20
##        417        426        429        430        430        430        429
## 2020-02-21 2020-02-22 2020-02-23 2020-02-24 2020-02-25 2020-02-26 2020-02-28
##        430        430         8        430        430        430        271
## 2020-02-29 2020-03-01 2020-03-02 2020-03-03 2020-03-04 2020-03-05 2020-03-06
##        314        273       114        430        313       117       377
## 2020-03-07 2020-03-08 2020-03-11 2020-03-12 2020-03-13 2020-03-14 2020-03-15
##        46         418       379        51        428       213       214
## 2020-03-18 2020-03-19 2020-03-20 2020-03-21 2020-03-22 2020-03-24 2020-03-25
##        371         58       356        72        428        429       429
## 2020-03-26 2020-03-27 2020-03-28 2020-03-29 2020-03-30 2020-03-31 2020-04-01
##        401        429       429       368        429       429       429
## [1] "How many observations for S_Leaf_area_cmsquared"
## 2020-02-14 2020-02-15 2020-02-16 2020-02-17 2020-02-18 2020-02-19 2020-02-20
##        408        426        429        430        430        430        429
## 2020-02-21 2020-02-22 2020-02-23 2020-02-24 2020-02-25 2020-02-26 2020-02-28
##        430        430         8        430        430        430        271
## 2020-02-29 2020-03-01 2020-03-02 2020-03-03 2020-03-04 2020-03-05 2020-03-06
##        314        273       114        430        313       117       377
## 2020-03-07 2020-03-08 2020-03-11 2020-03-12 2020-03-13 2020-03-14 2020-03-15
##        46         418       379        51        428       213       214
## 2020-03-18 2020-03-19 2020-03-20 2020-03-21 2020-03-22 2020-03-24 2020-03-25
##        371         58       356        72        428        429       429
## 2020-03-26 2020-03-27 2020-03-28 2020-03-29 2020-03-30 2020-03-31 2020-04-01
##        401        429       429       368        429       429       429

```

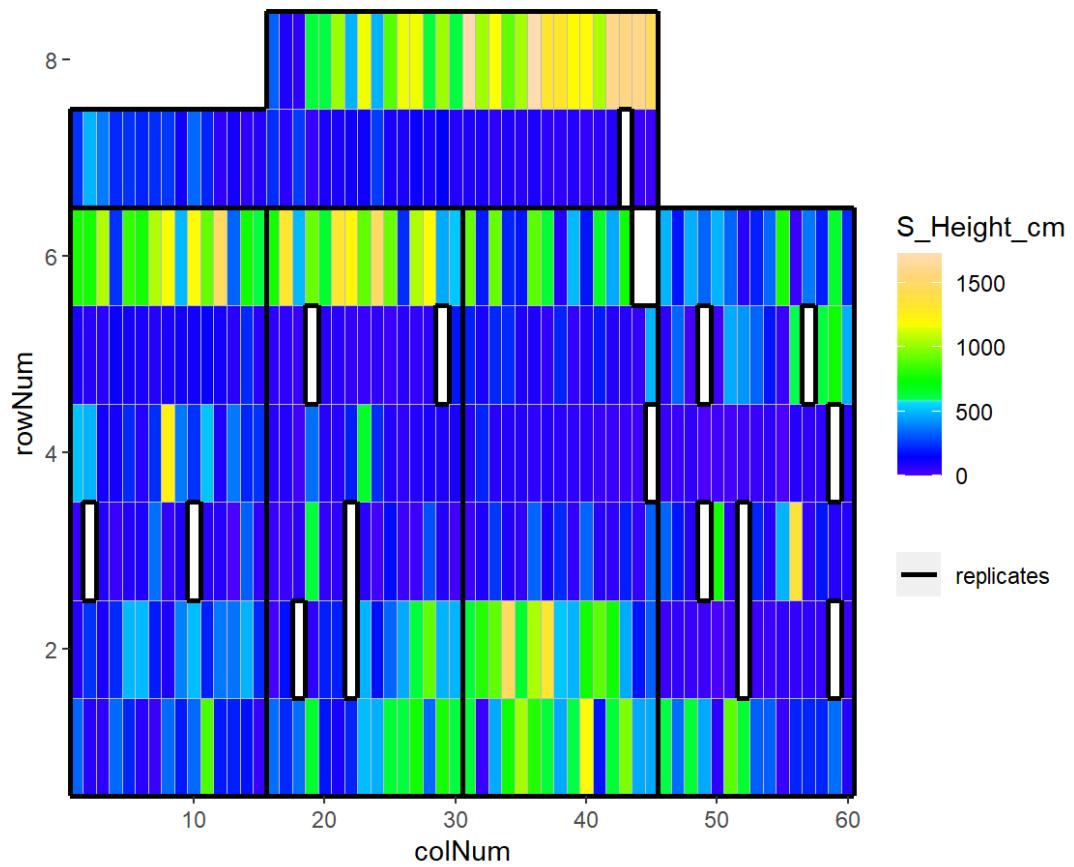
Check the heatmap of the raw data per time point

```

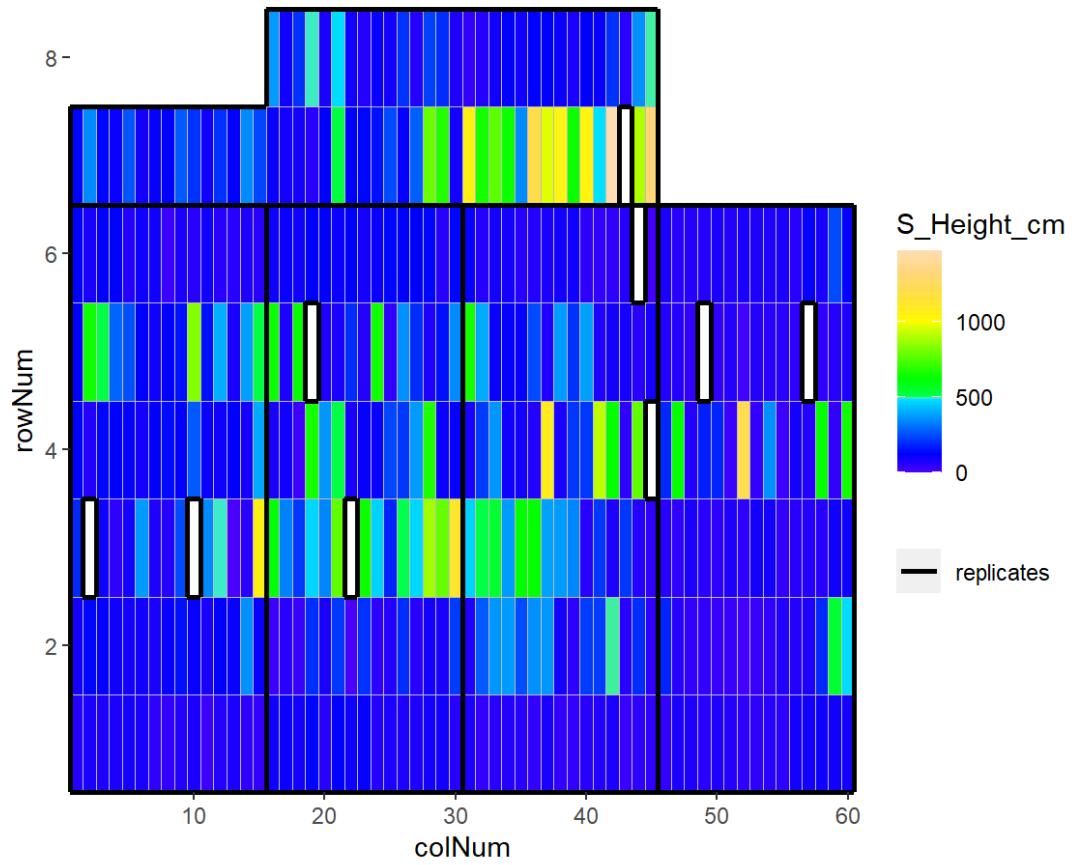
for (trait_name in traits) {
  for (tp in 1:length(num_timepoints$timeNumber)) {
    plot(timePoint_S,
      plotType = "layout",
      timePoints = tp,
      traits = trait_name)
  }
}

```

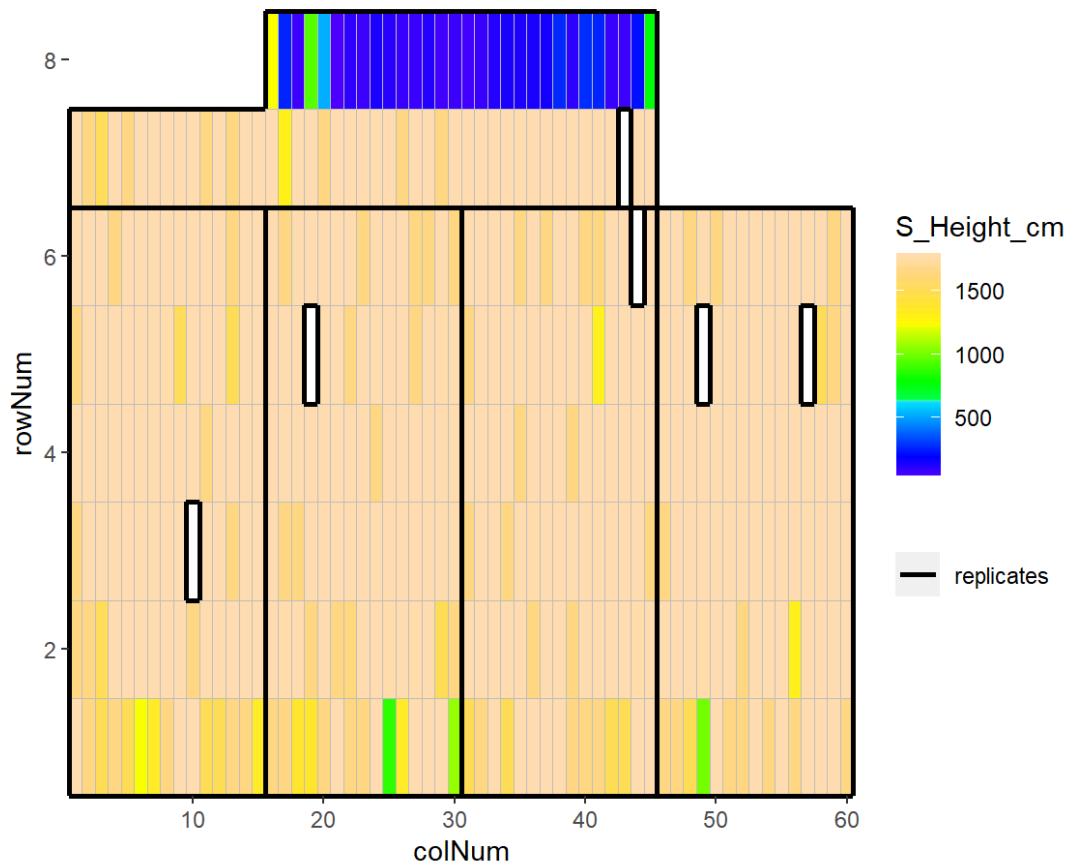
## EPPN2020\_M3P - 2020-02-14



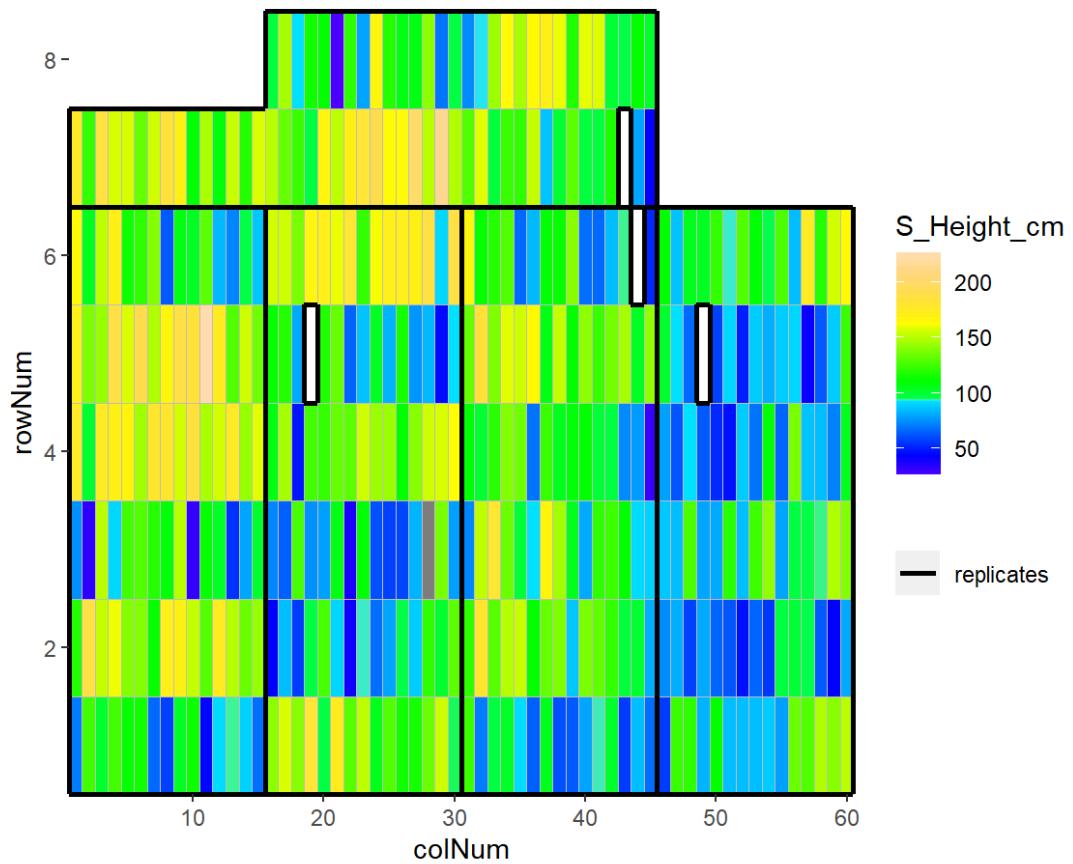
## EPPN2020\_M3P - 2020-02-15



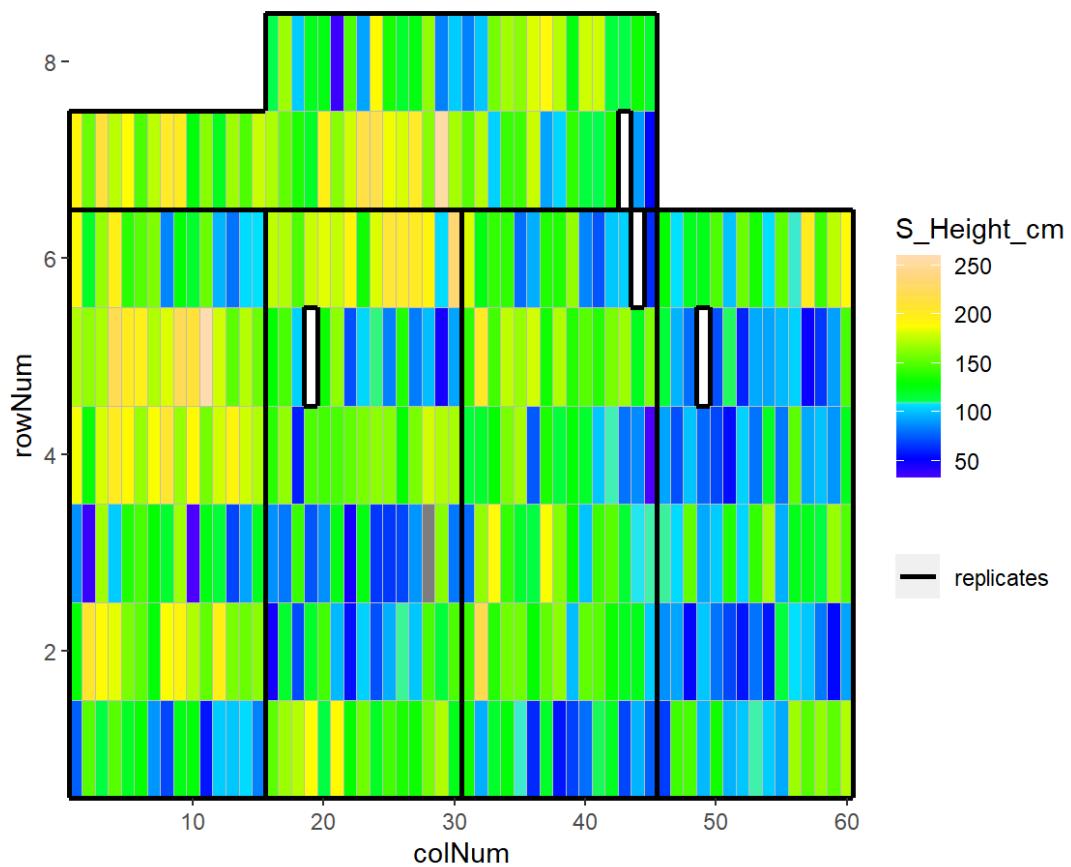
## EPPN2020\_M3P - 2020-02-16



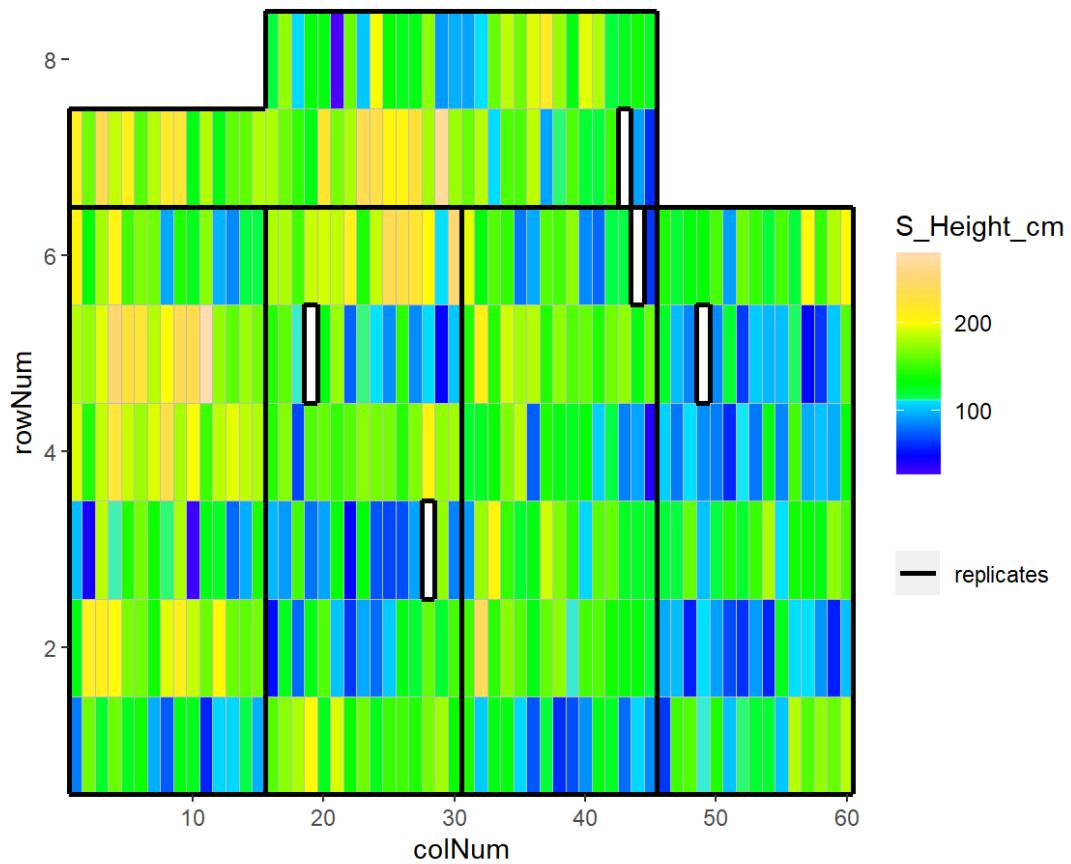
## EPPN2020\_M3P - 2020-02-17



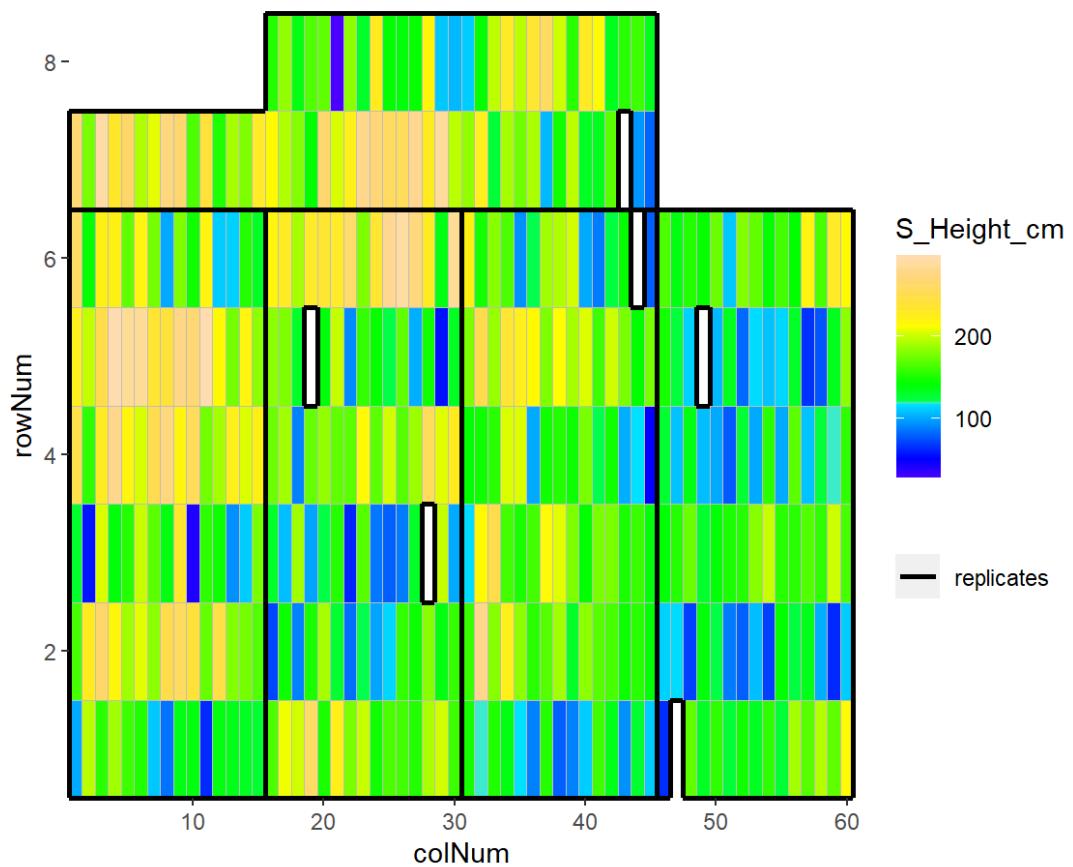
## EPPN2020\_M3P - 2020-02-18



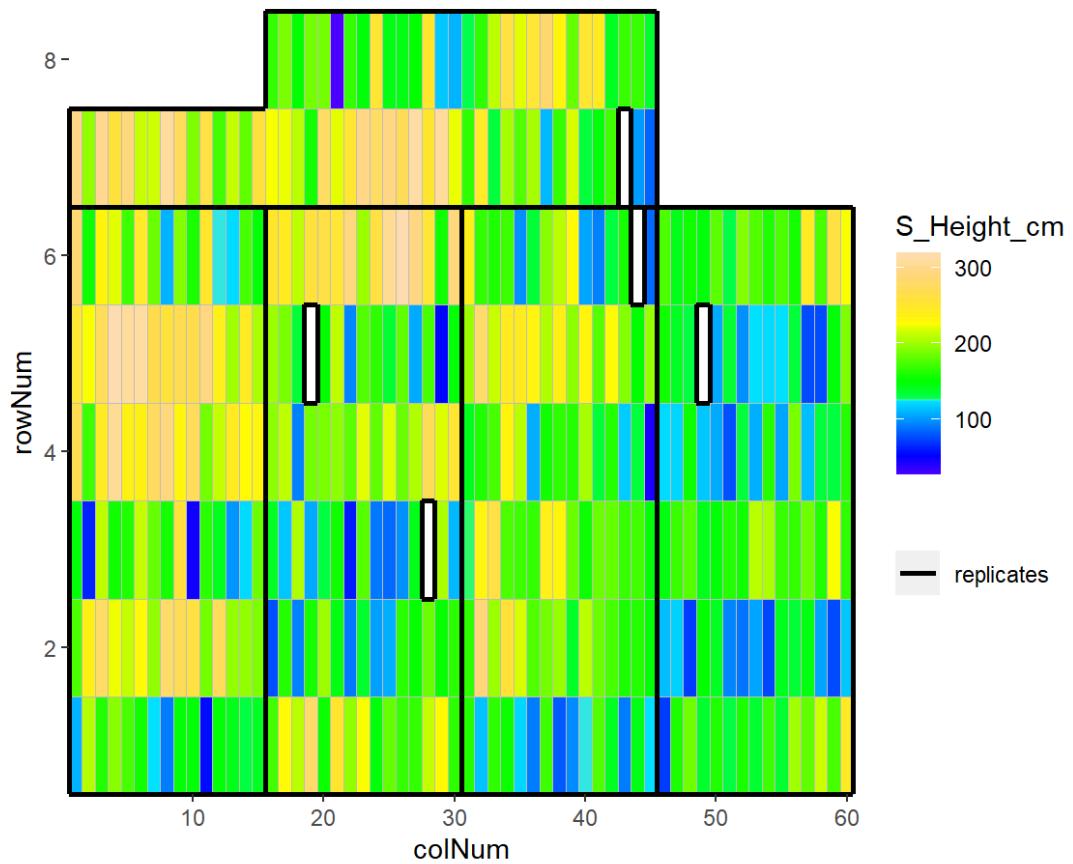
## EPPN2020\_M3P - 2020-02-19



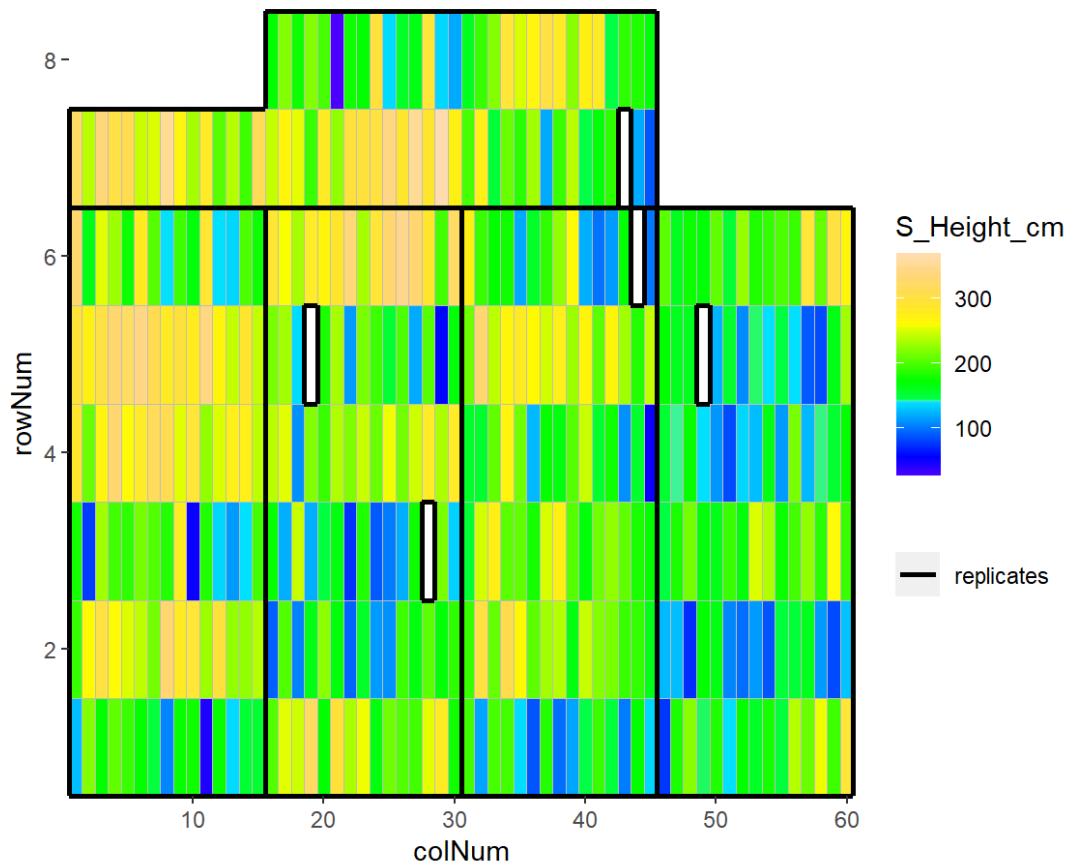
## EPPN2020\_M3P - 2020-02-20



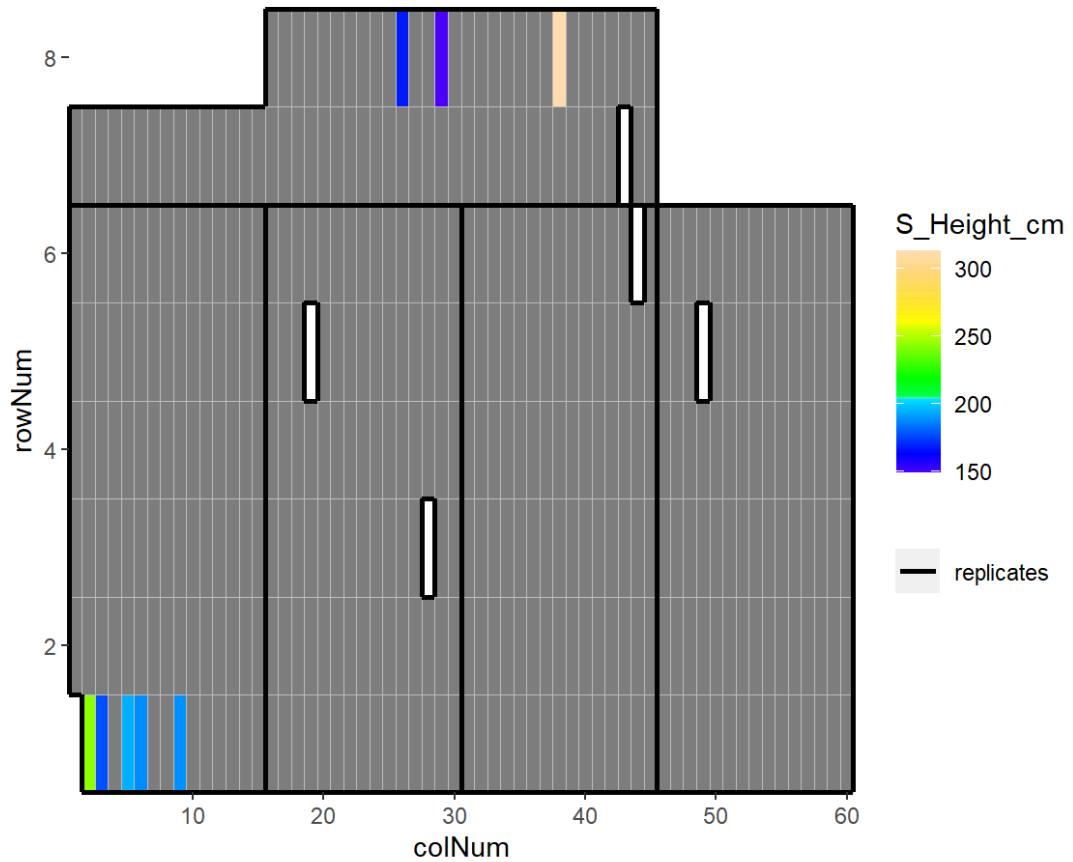
## EPPN2020\_M3P - 2020-02-21



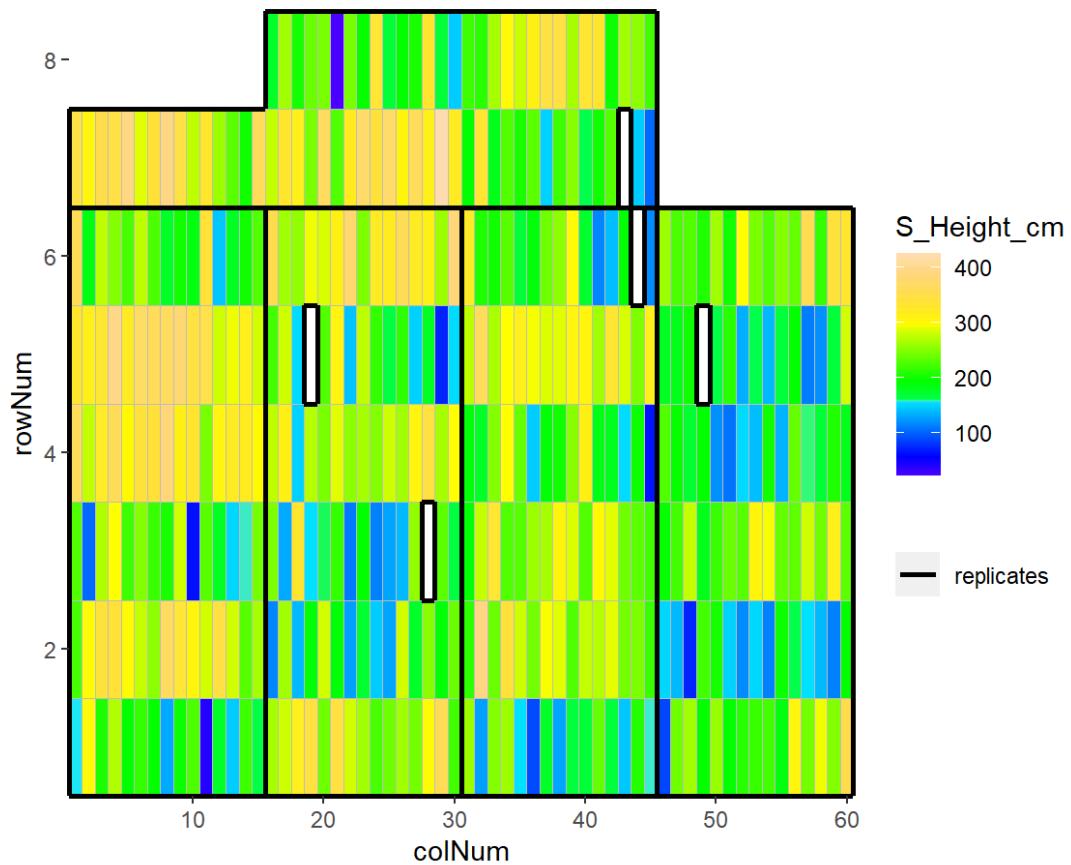
## EPPN2020\_M3P - 2020-02-22



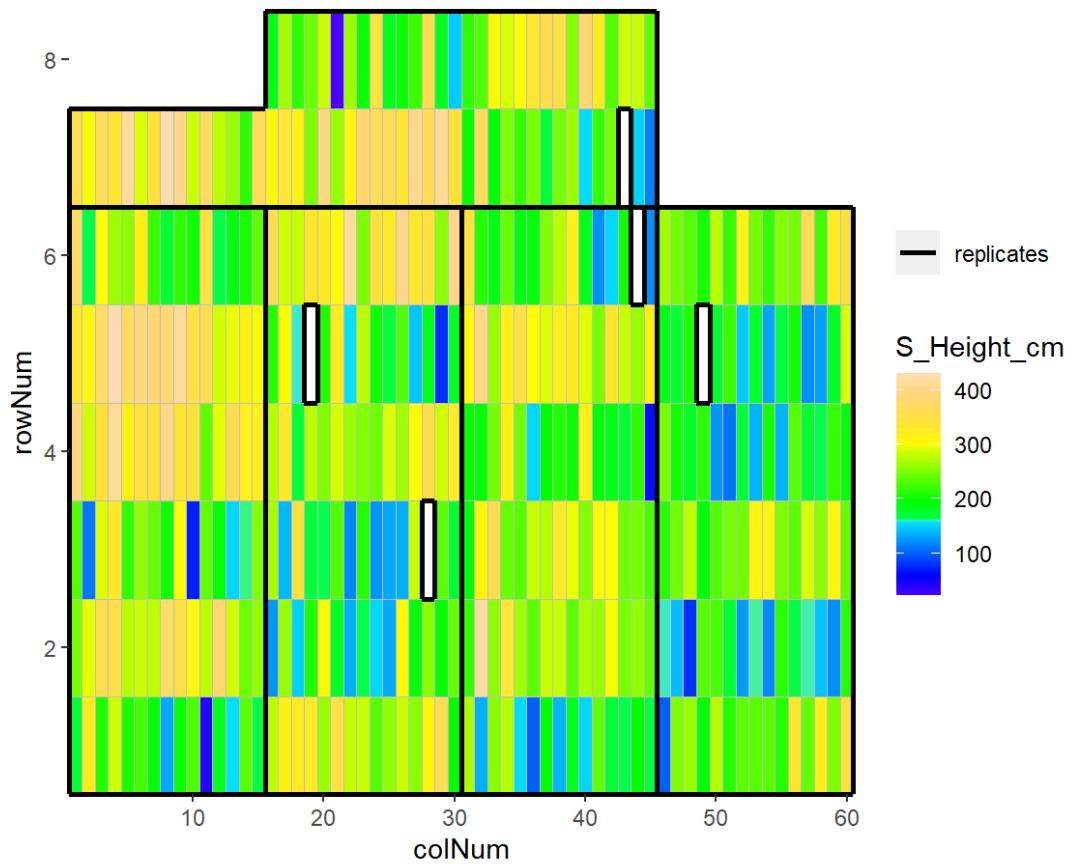
## EPPN2020\_M3P - 2020-02-23



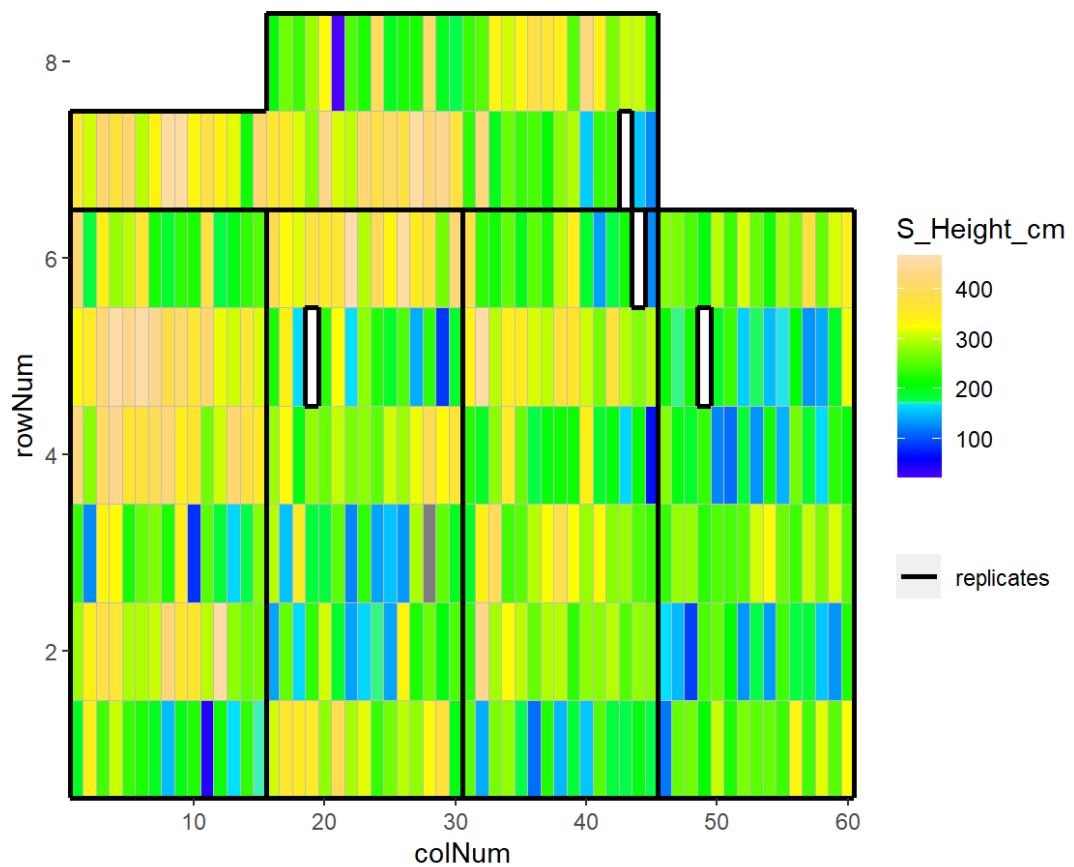
## EPPN2020\_M3P - 2020-02-24



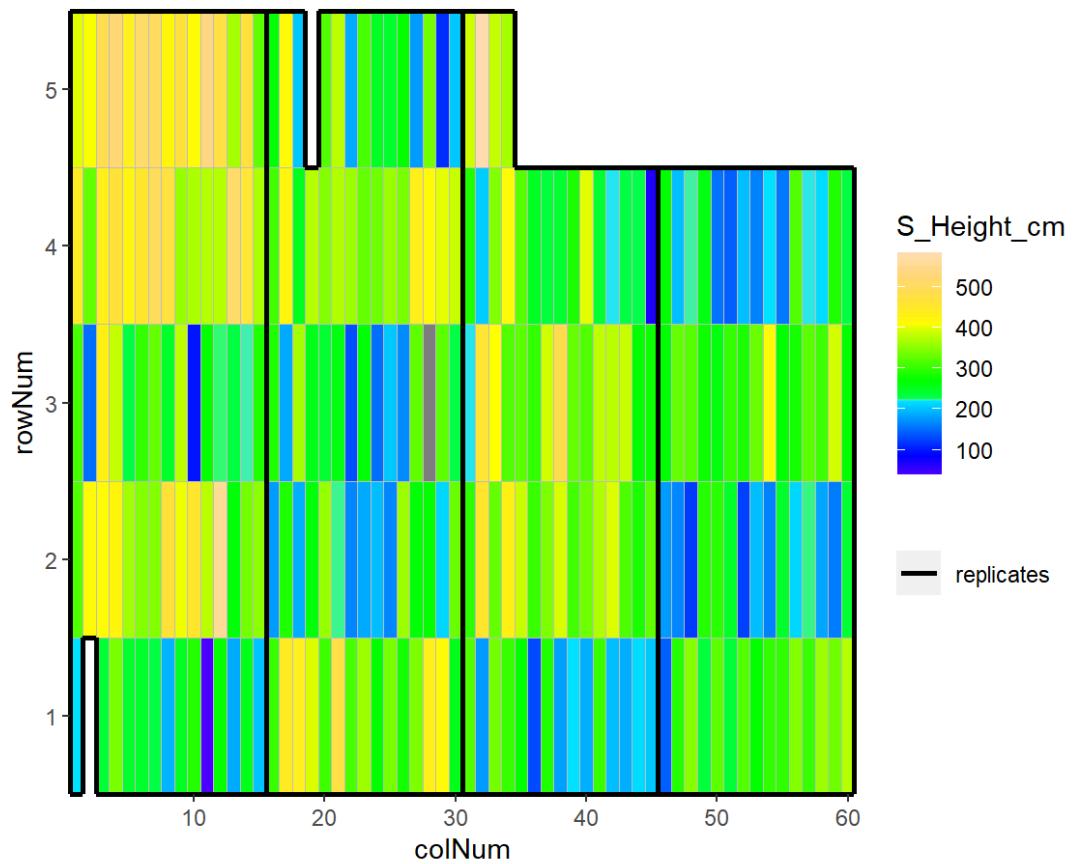
## EPPN2020\_M3P - 2020-02-25



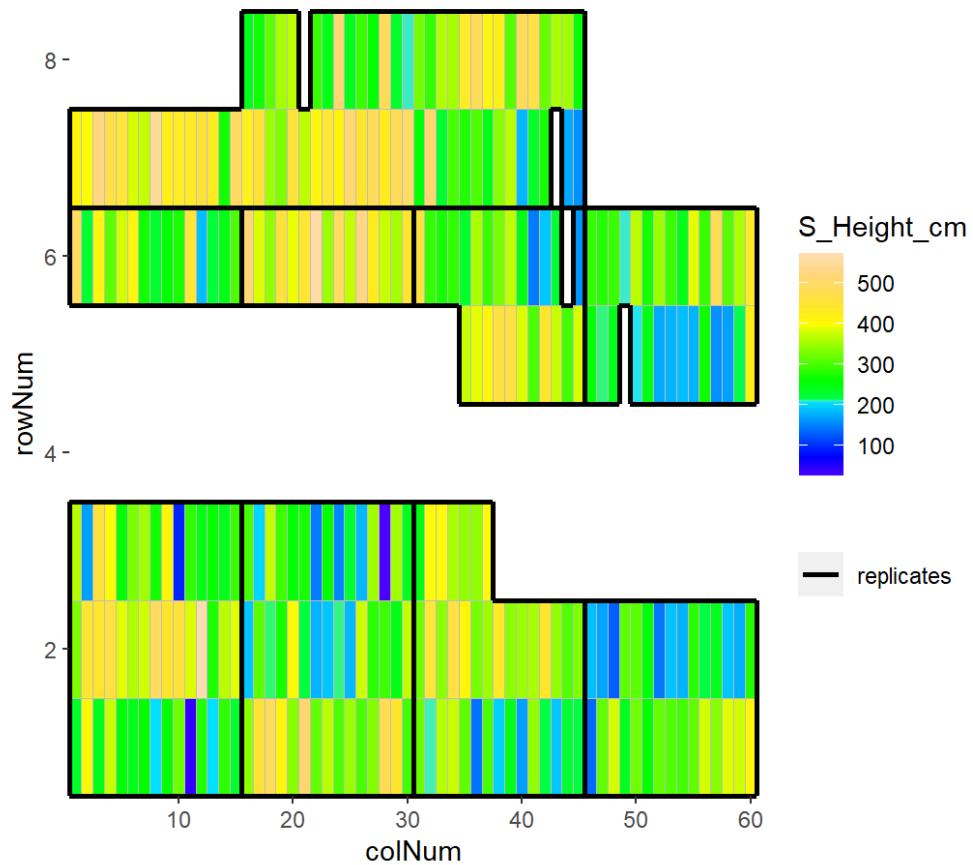
## EPPN2020\_M3P - 2020-02-26



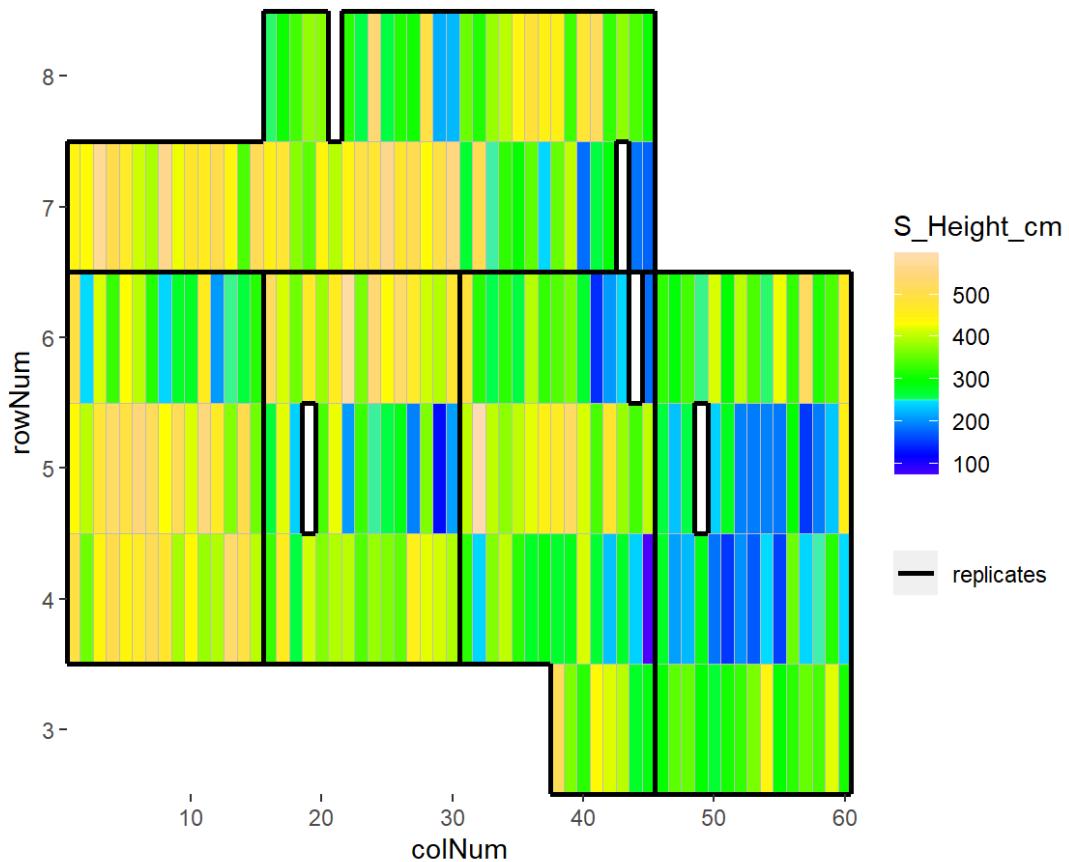
## EPPN2020\_M3P - 2020-02-28



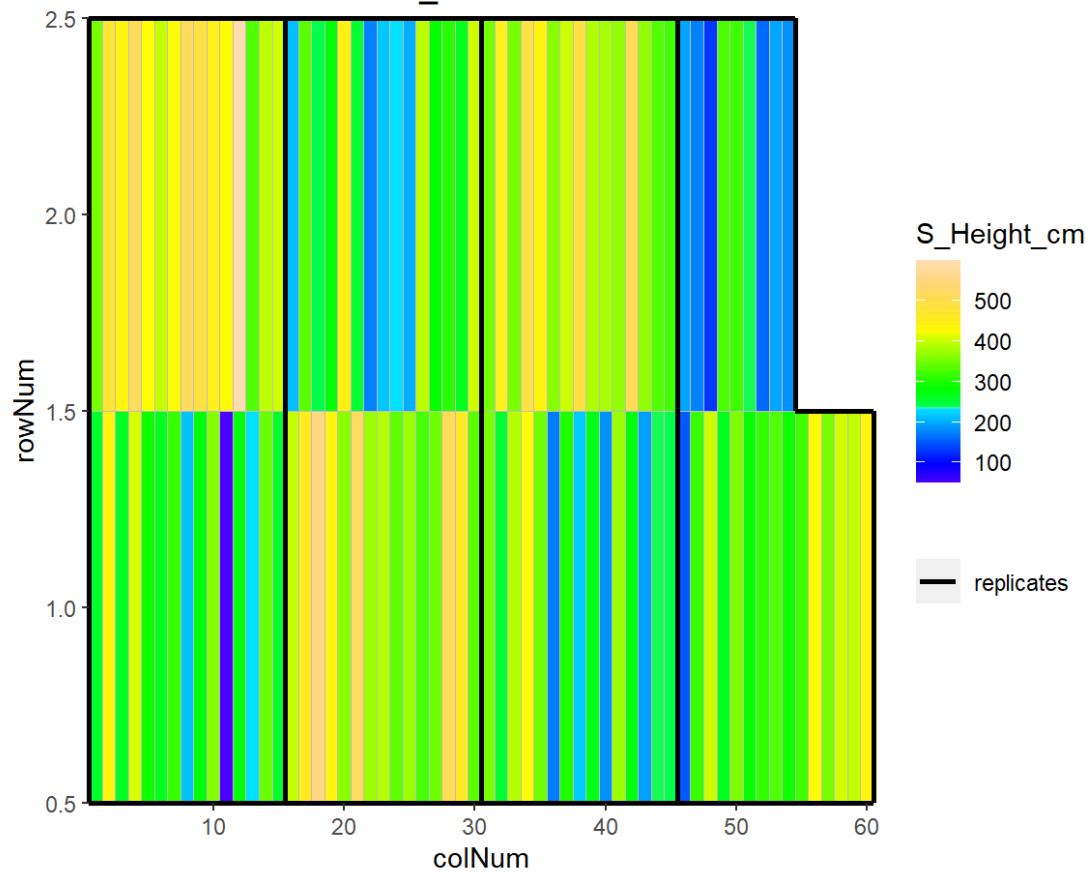
## EPPN2020\_M3P - 2020-02-29



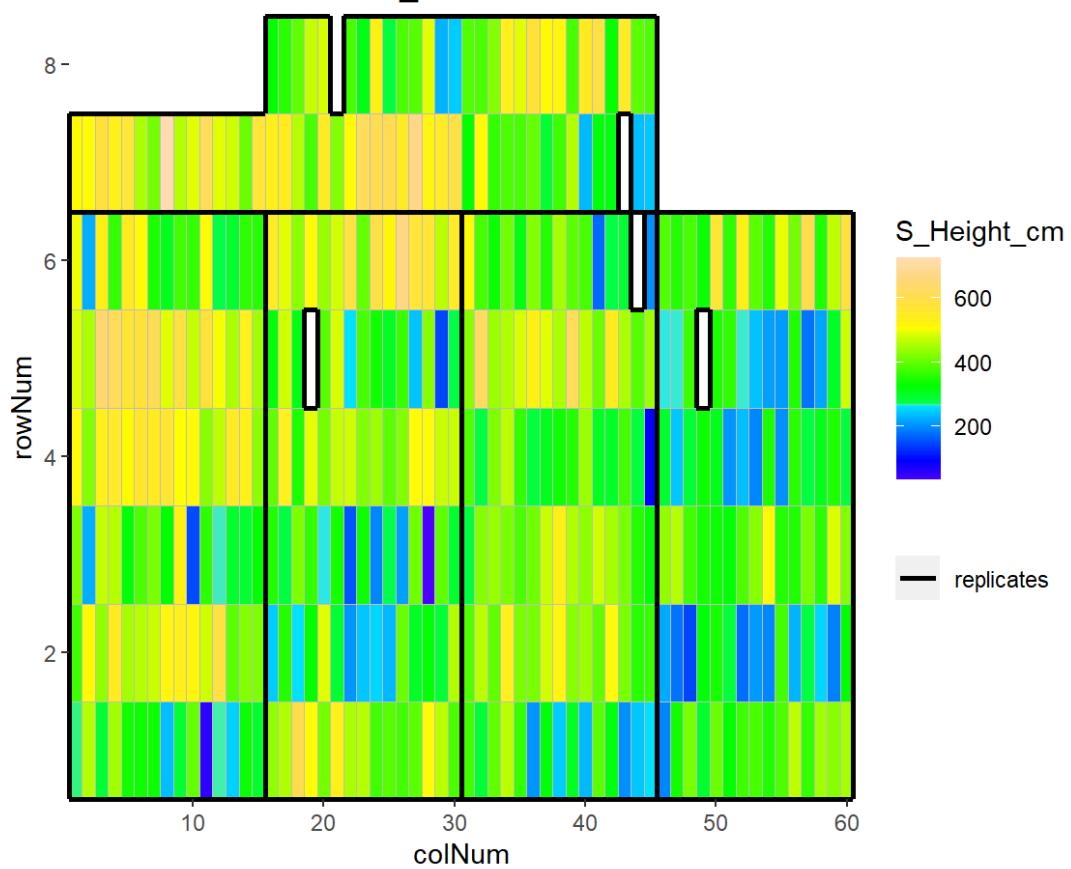
## EPPN2020\_M3P - 2020-03-01



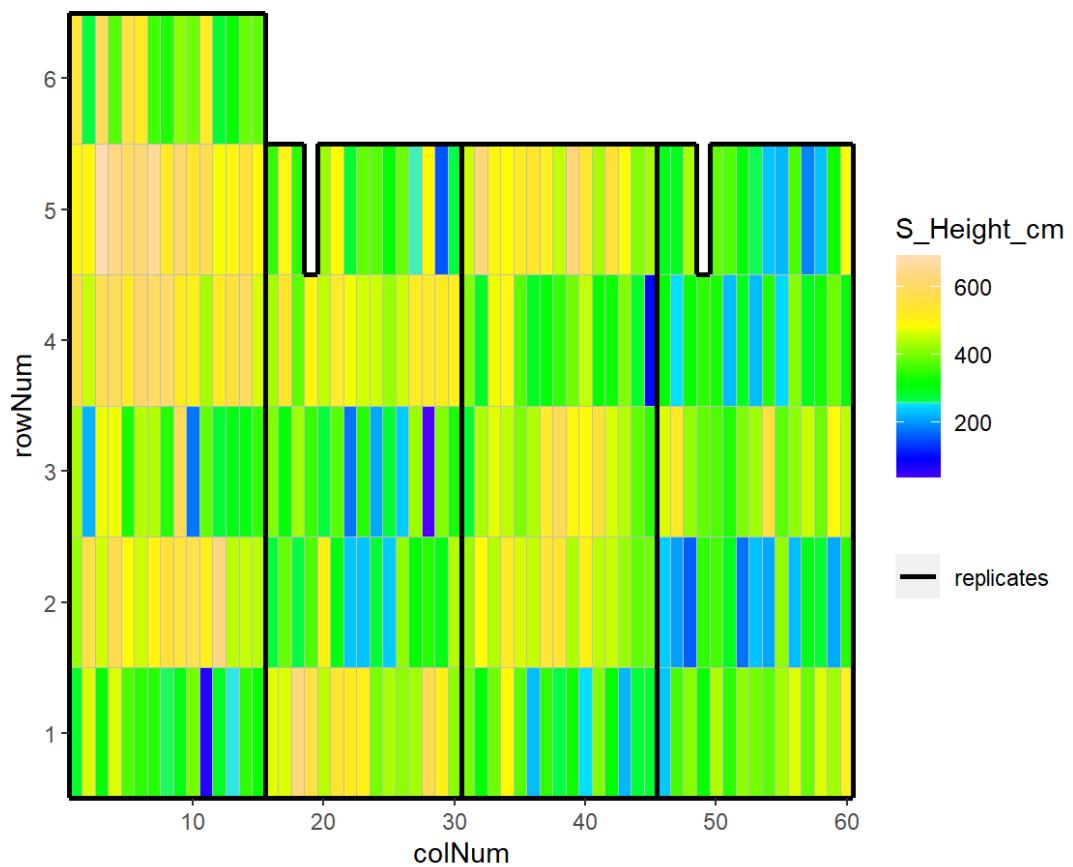
## EPPN2020\_M3P - 2020-03-02



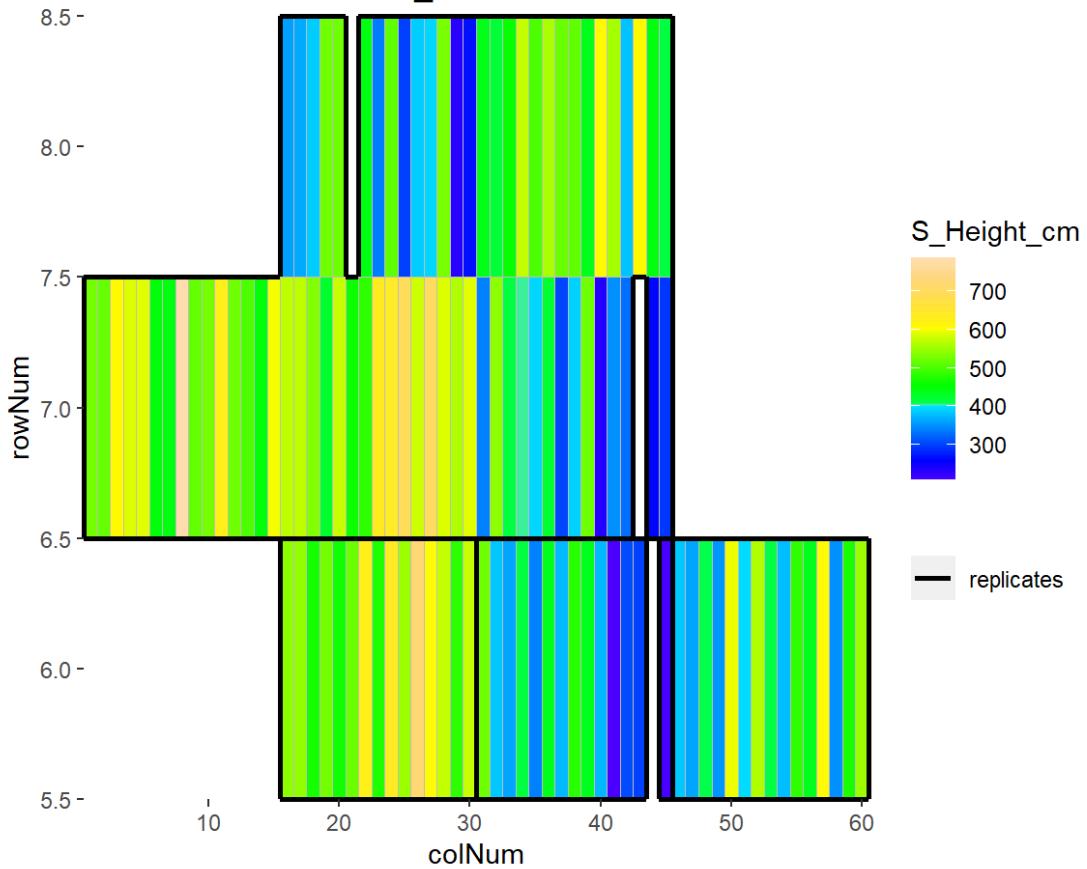
## EPPN2020\_M3P - 2020-03-03

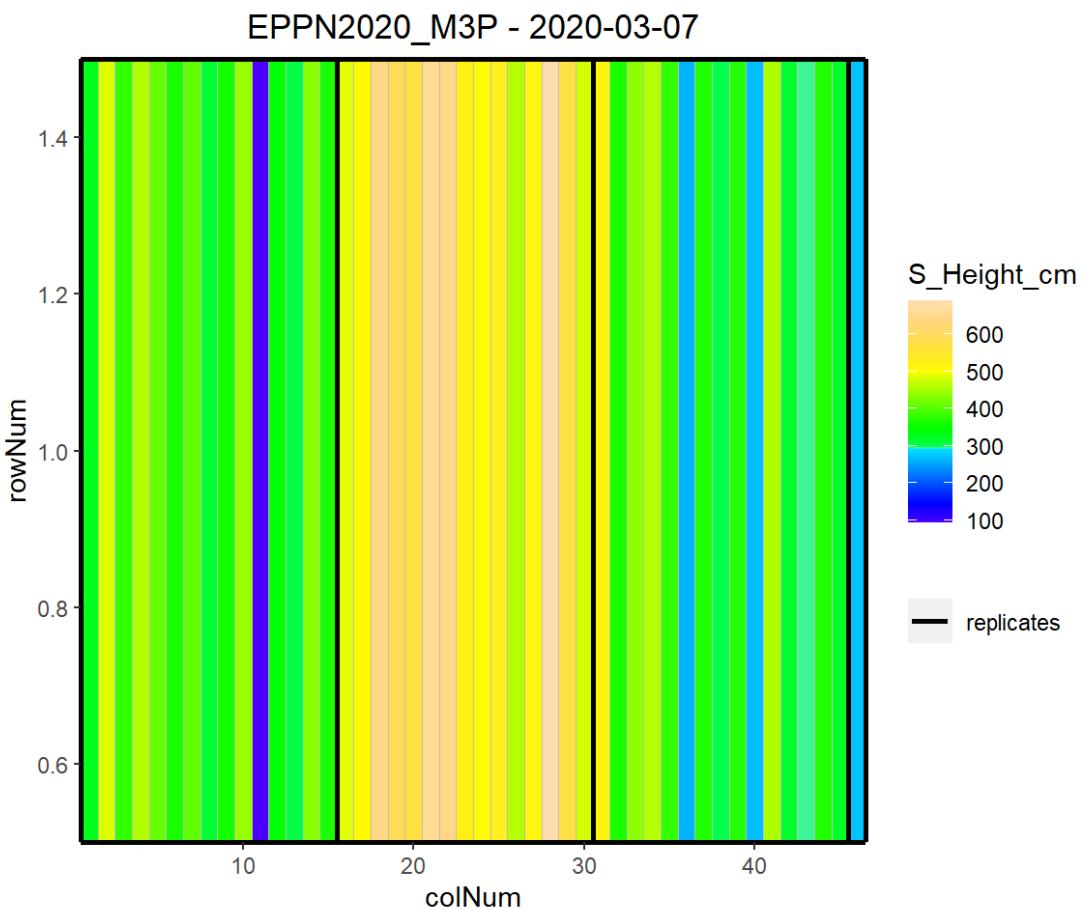
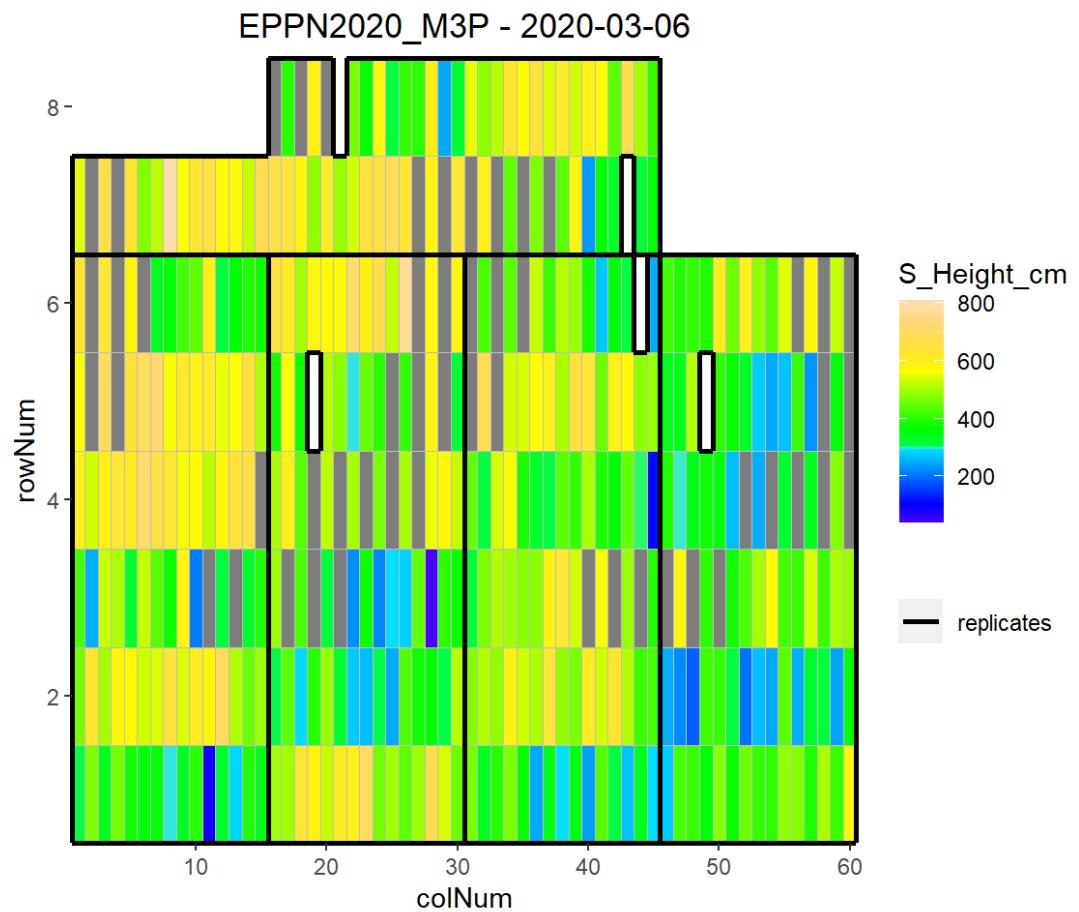


## EPPN2020\_M3P - 2020-03-04

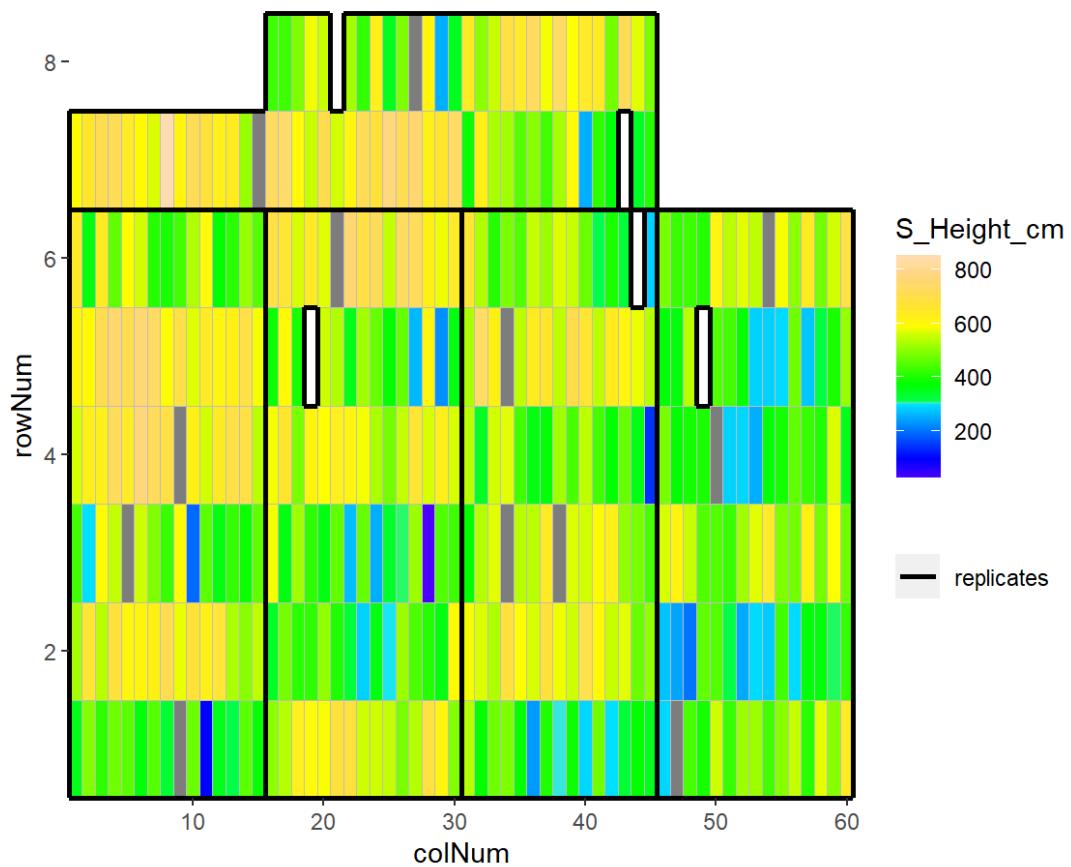


## EPPN2020\_M3P - 2020-03-05

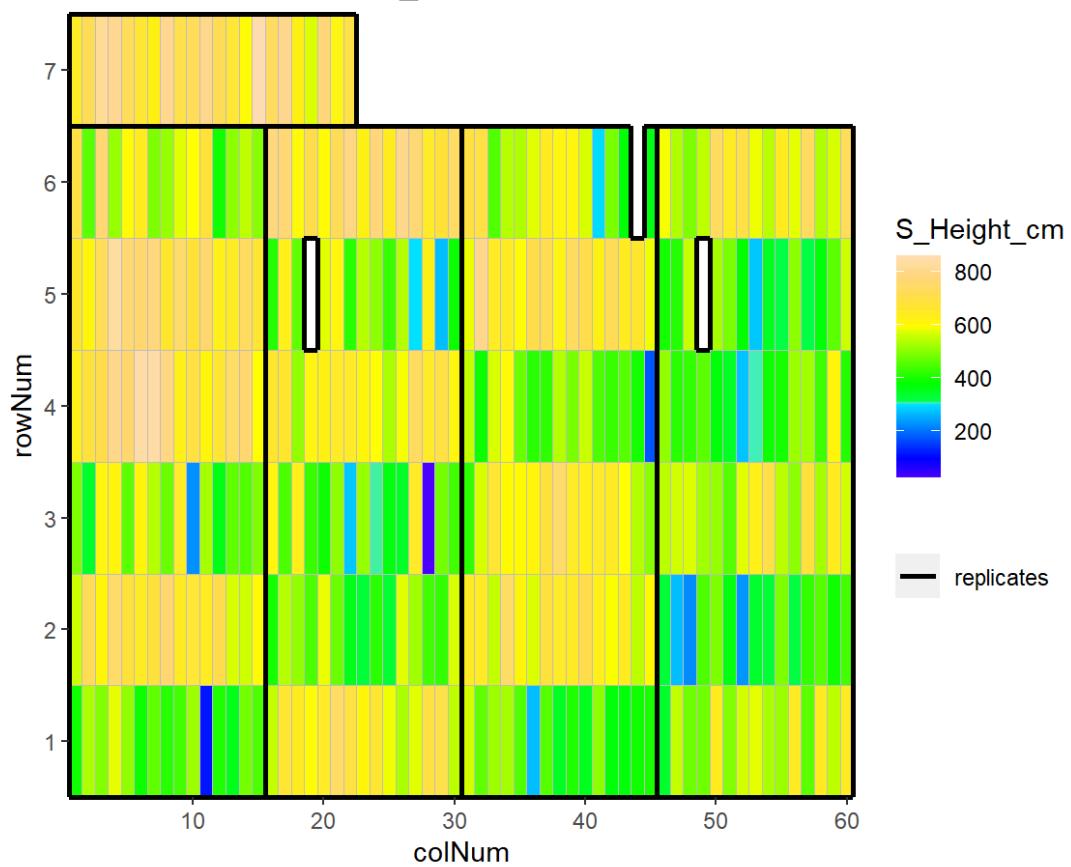


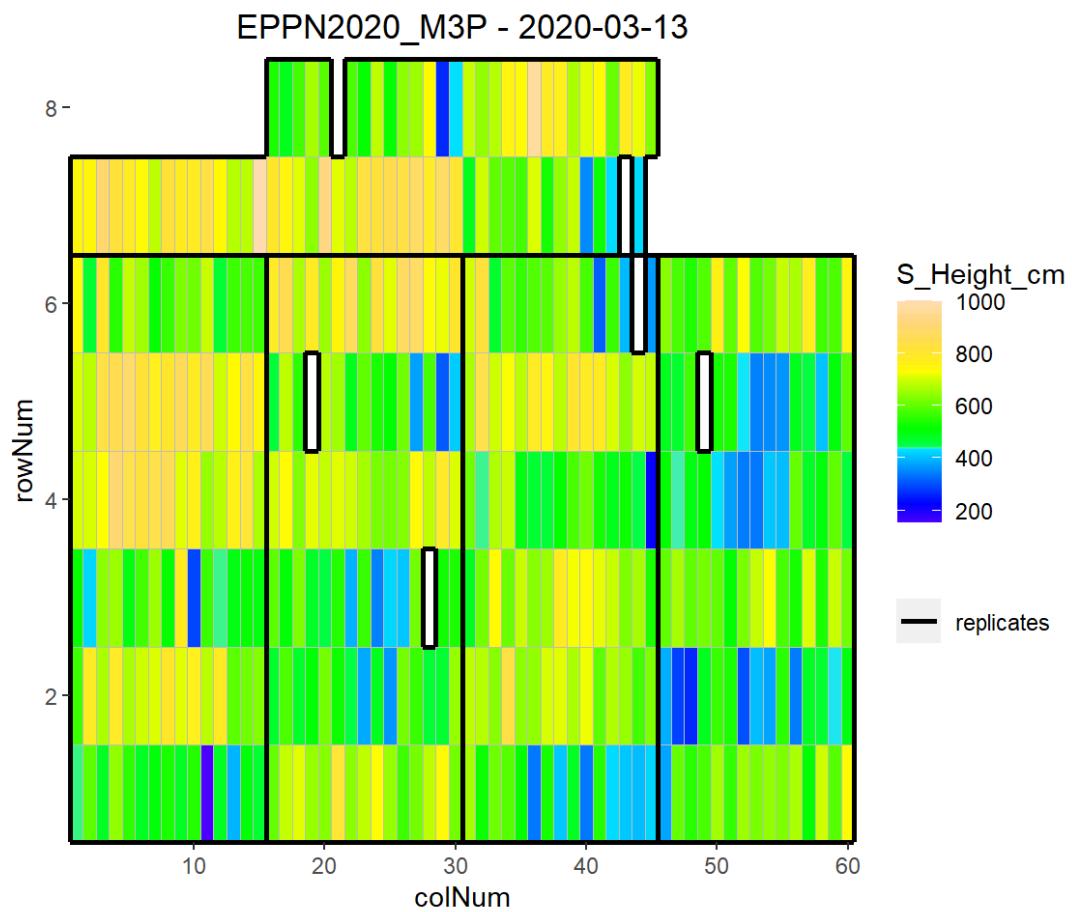
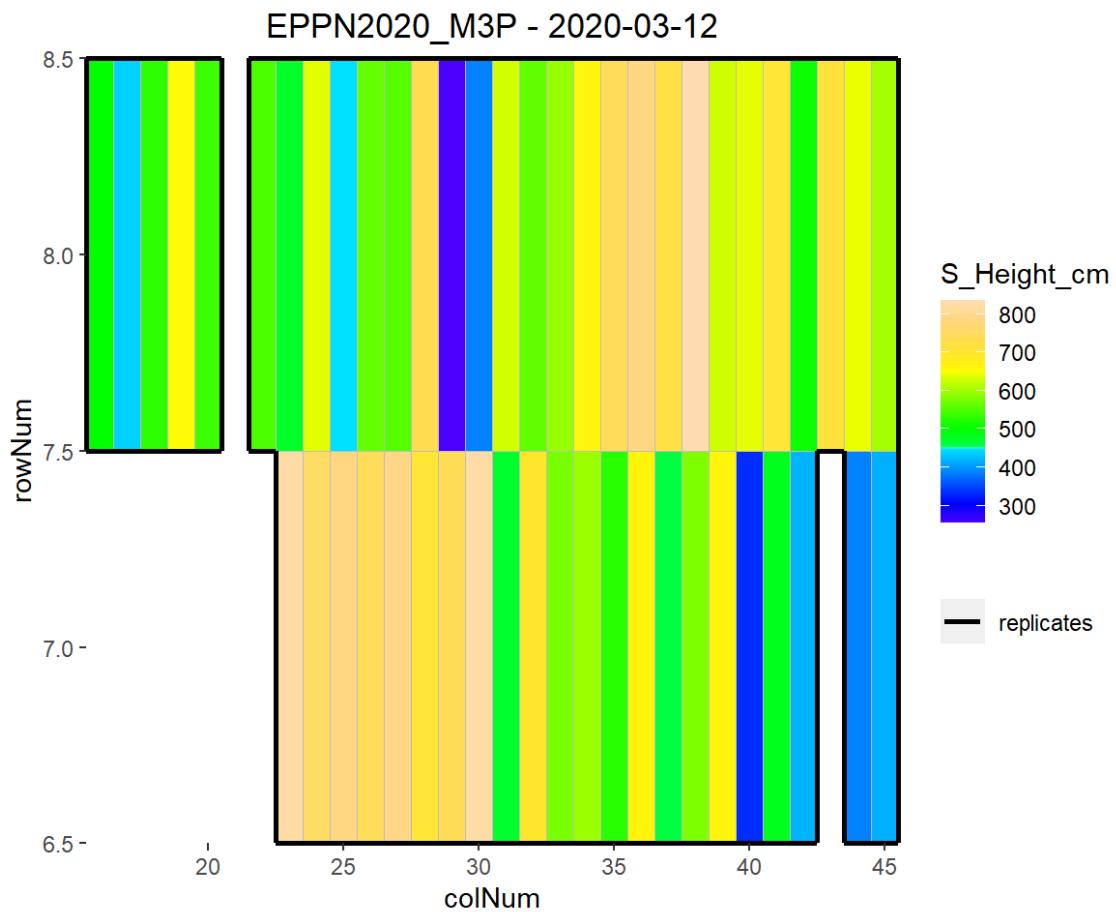


## EPPN2020\_M3P - 2020-03-08

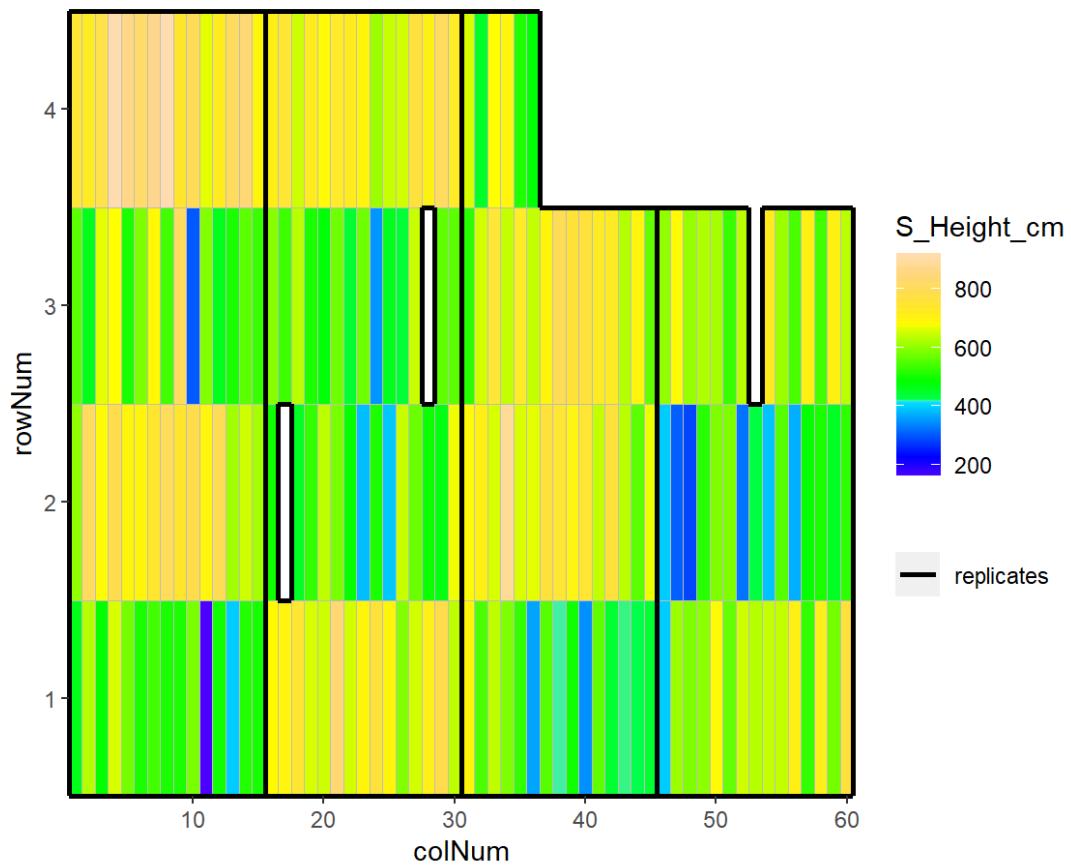


## EPPN2020\_M3P - 2020-03-11

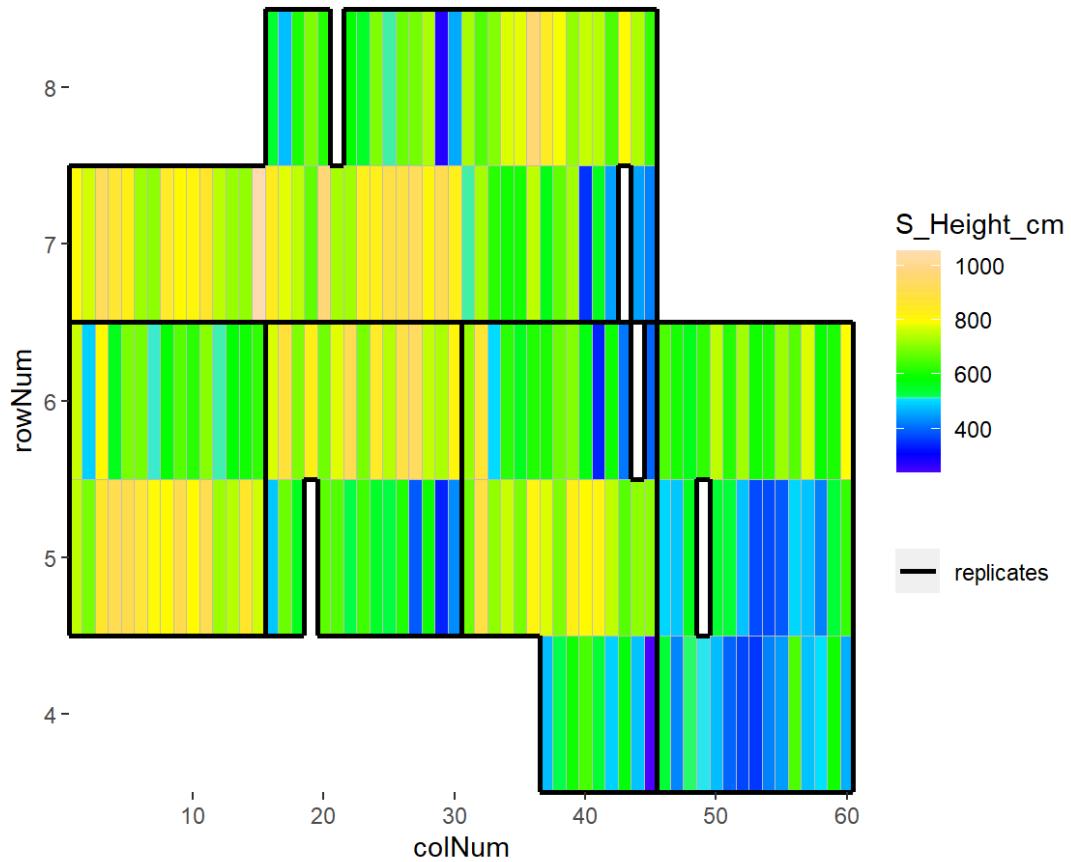




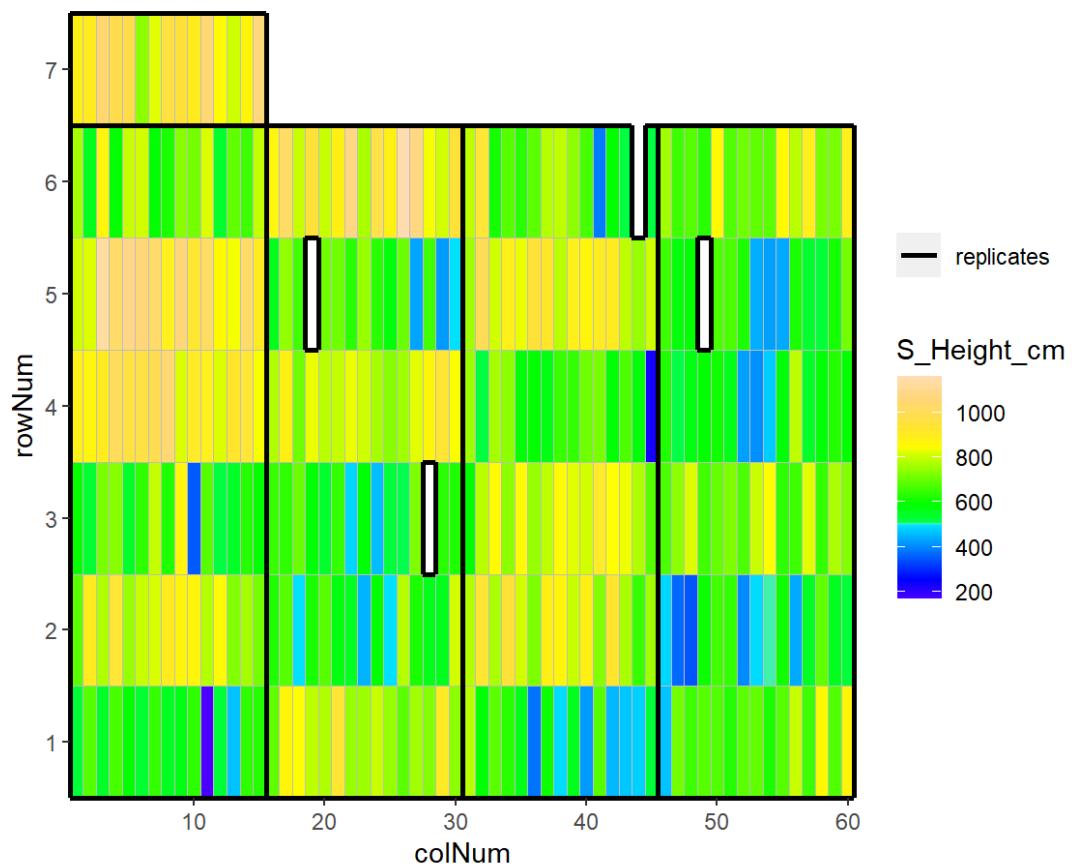
## EPPN2020\_M3P - 2020-03-14



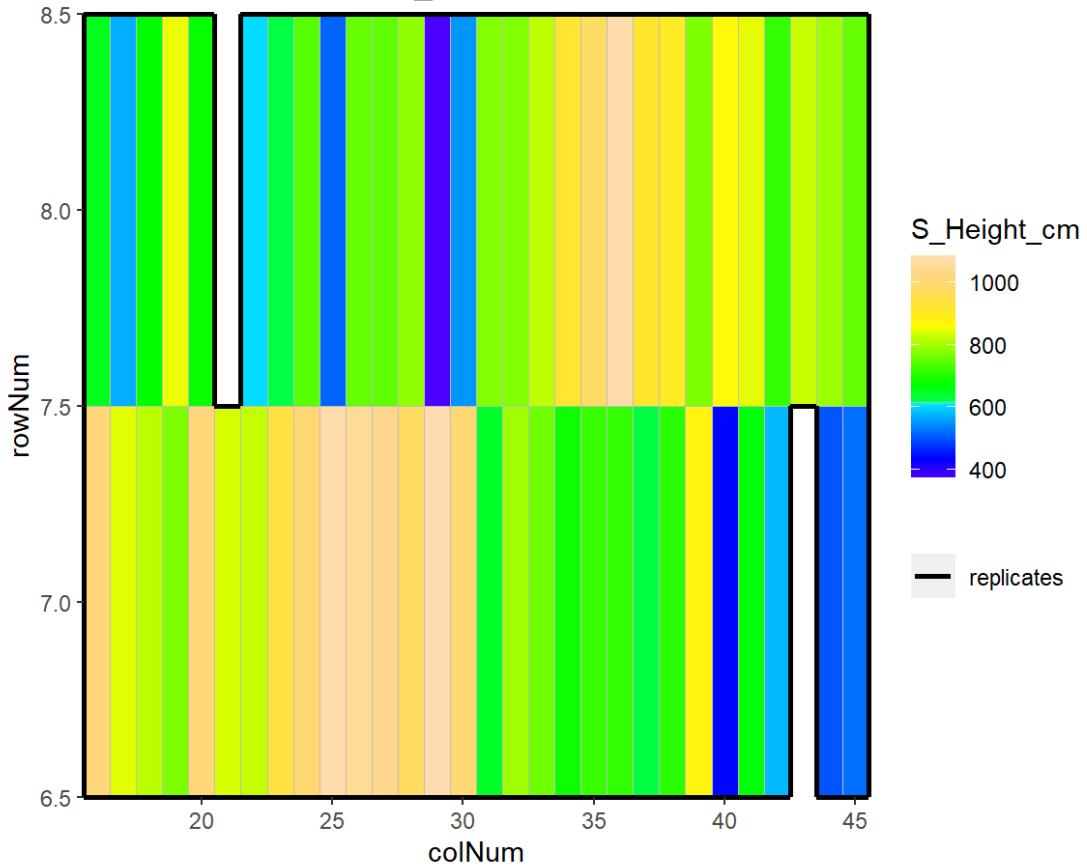
## EPPN2020\_M3P - 2020-03-15



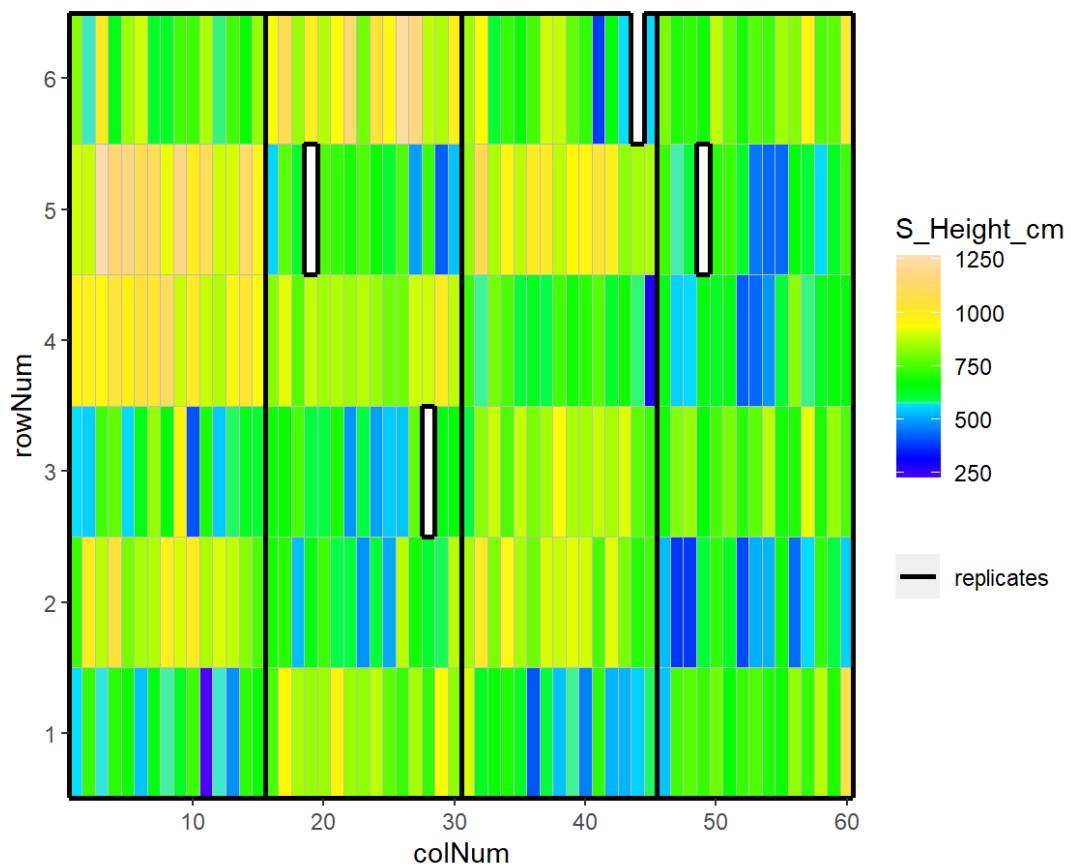
## EPPN2020\_M3P - 2020-03-18



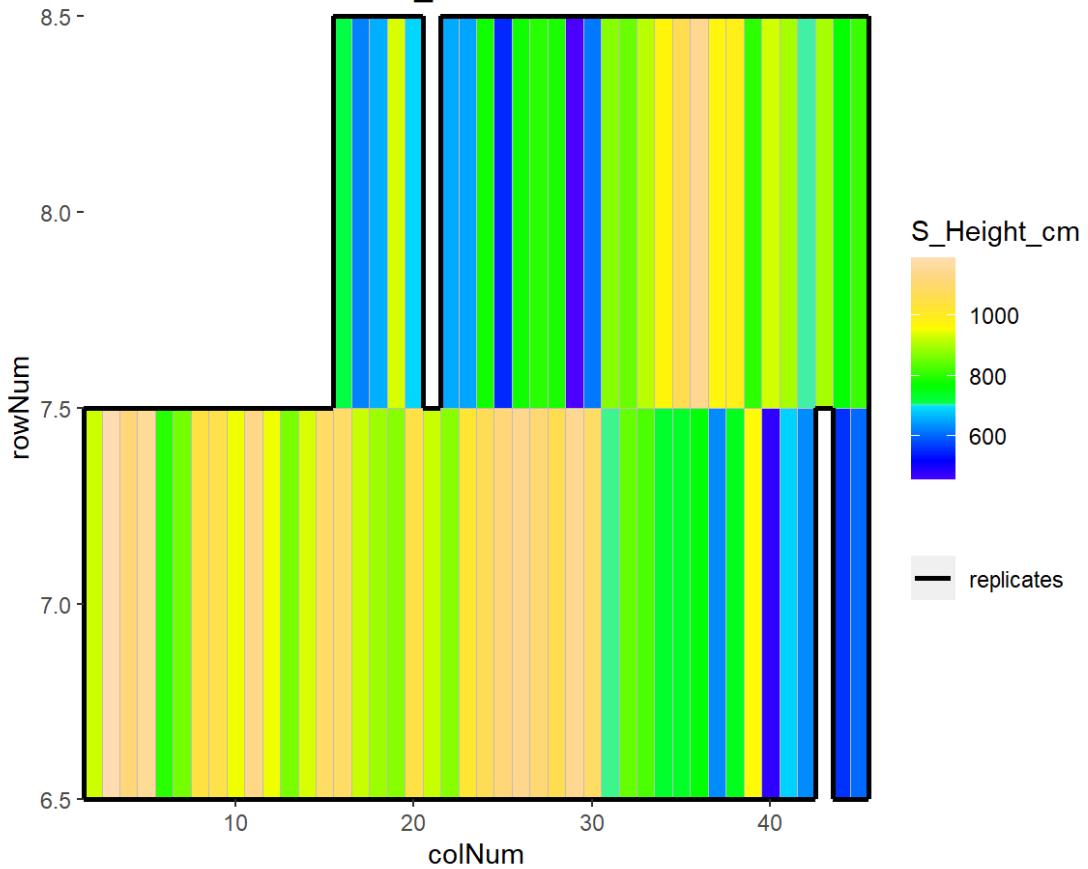
## EPPN2020\_M3P - 2020-03-19



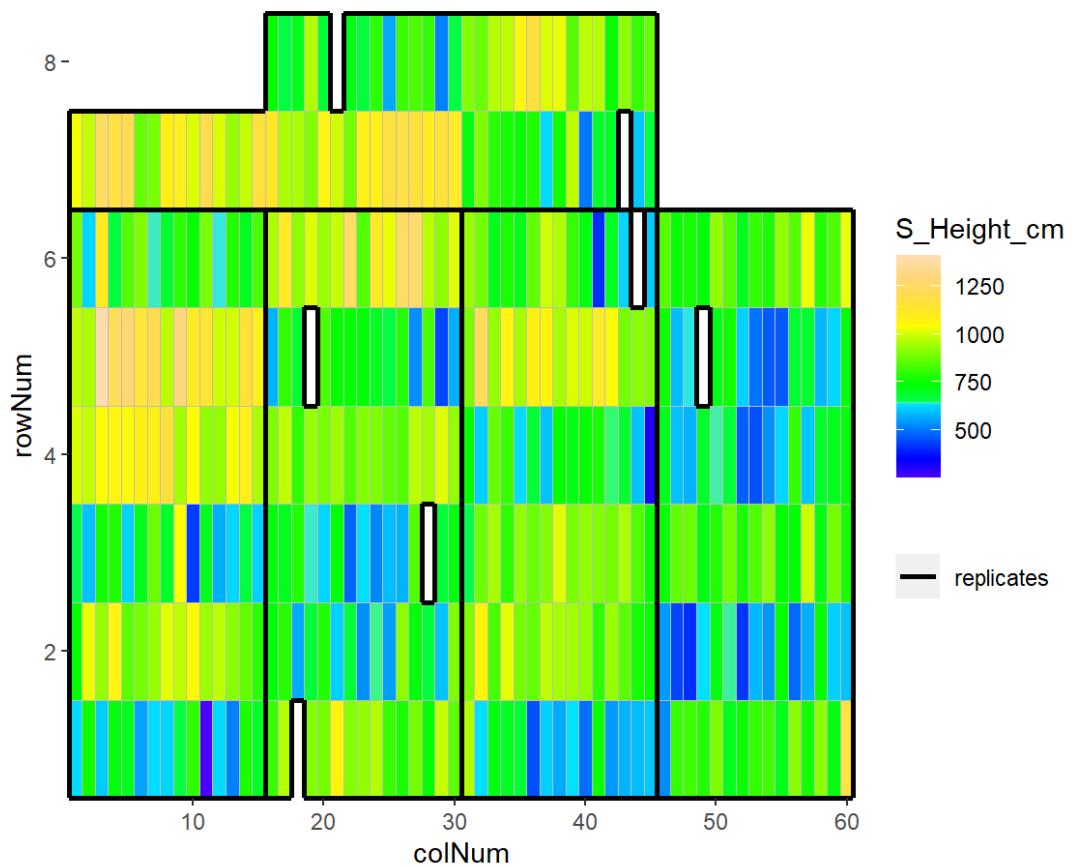
## EPPN2020\_M3P - 2020-03-20



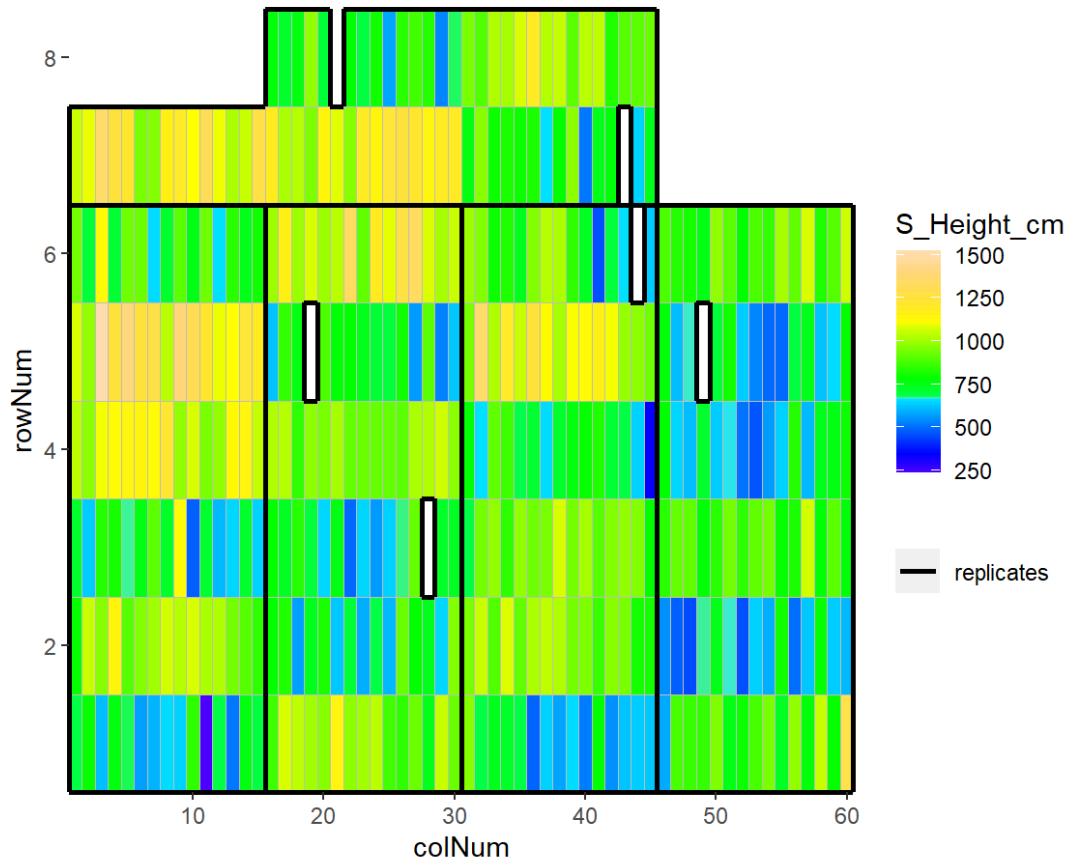
## EPPN2020\_M3P - 2020-03-21



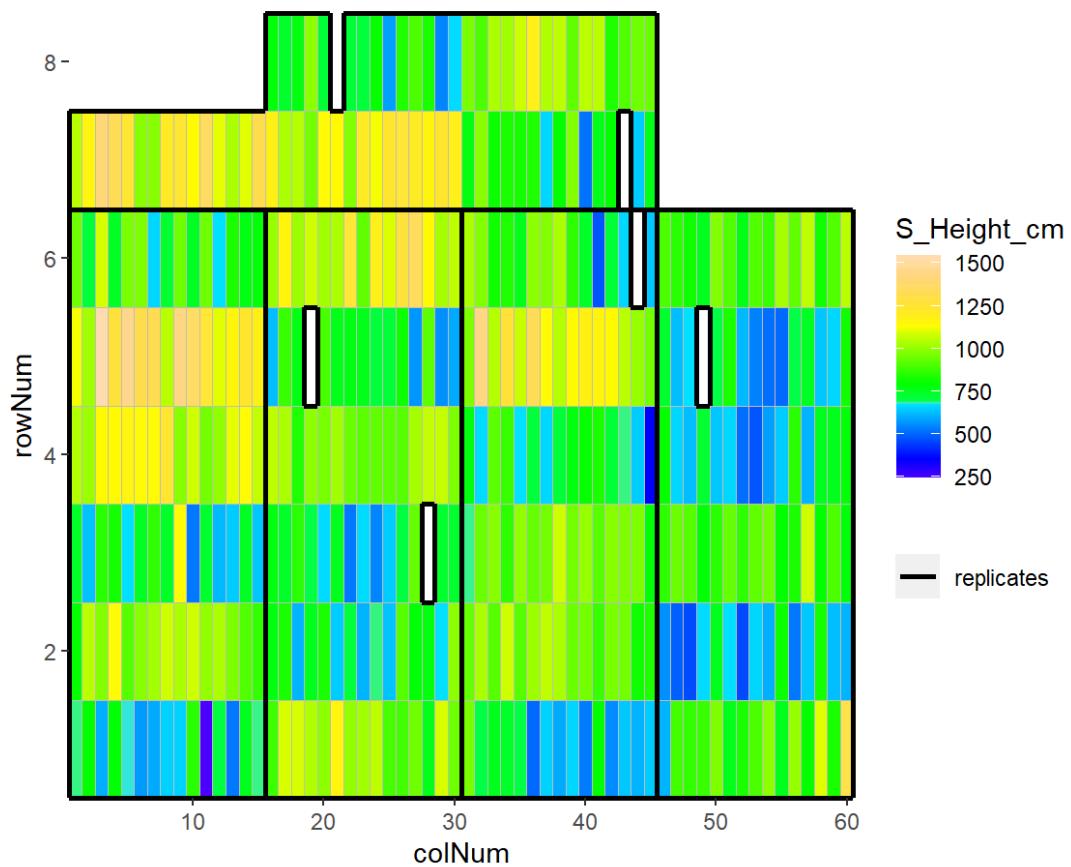
## EPPN2020\_M3P - 2020-03-22



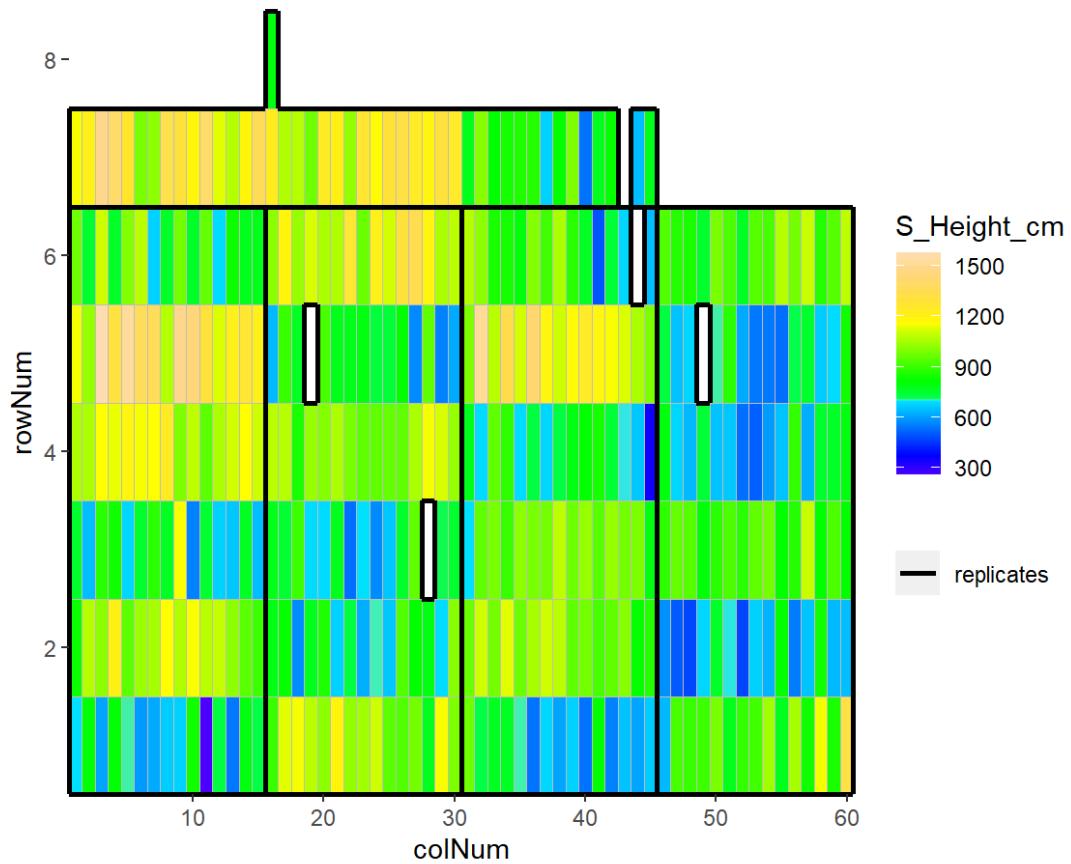
## EPPN2020\_M3P - 2020-03-24



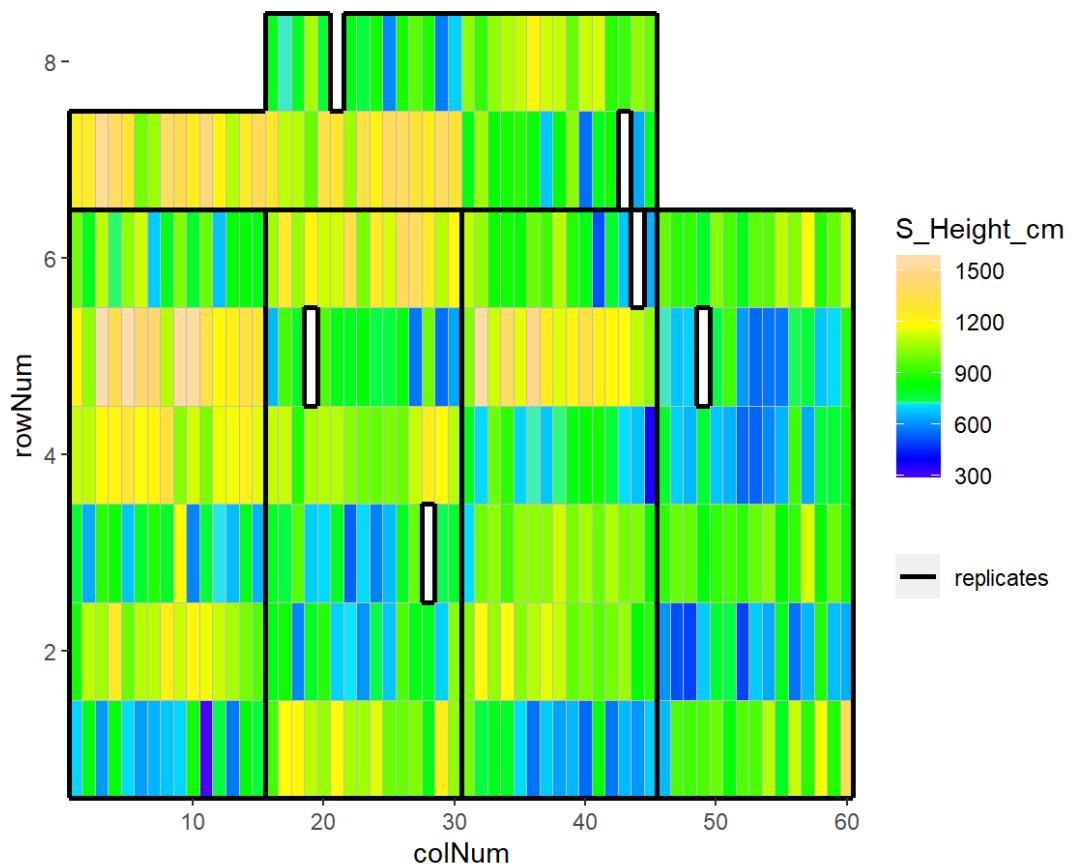
## EPPN2020\_M3P - 2020-03-25



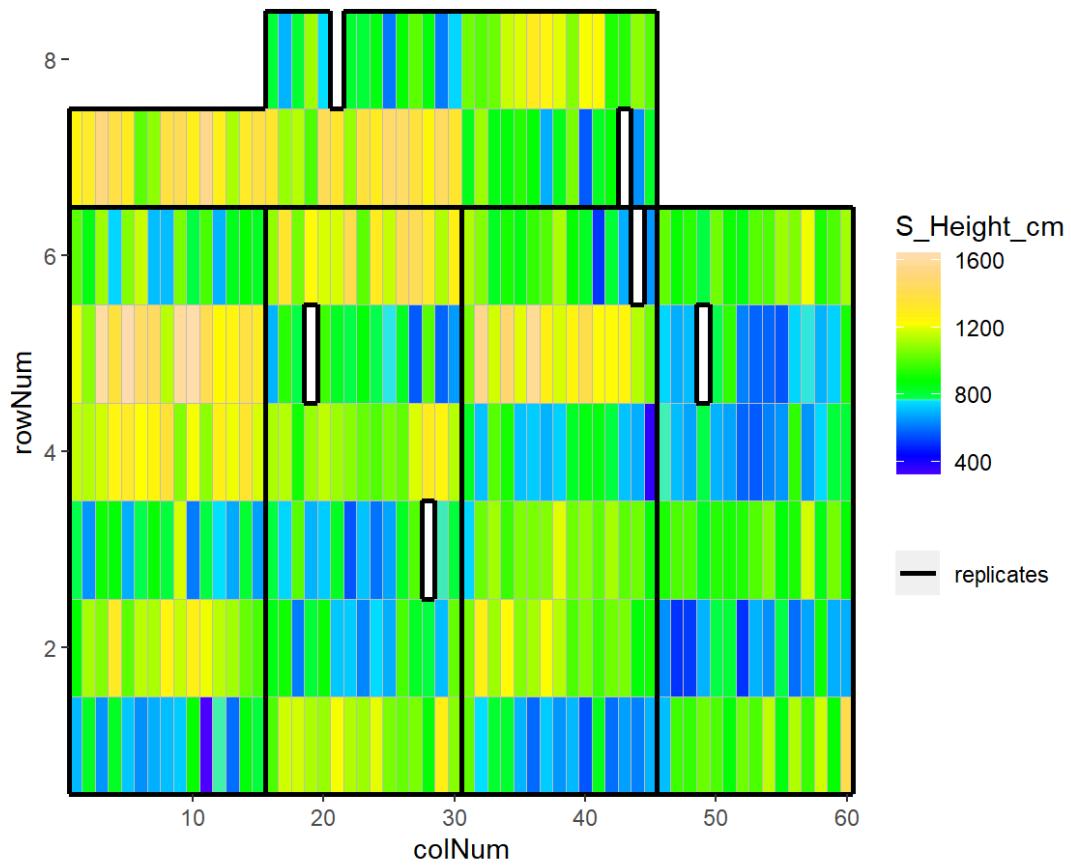
## EPPN2020\_M3P - 2020-03-26



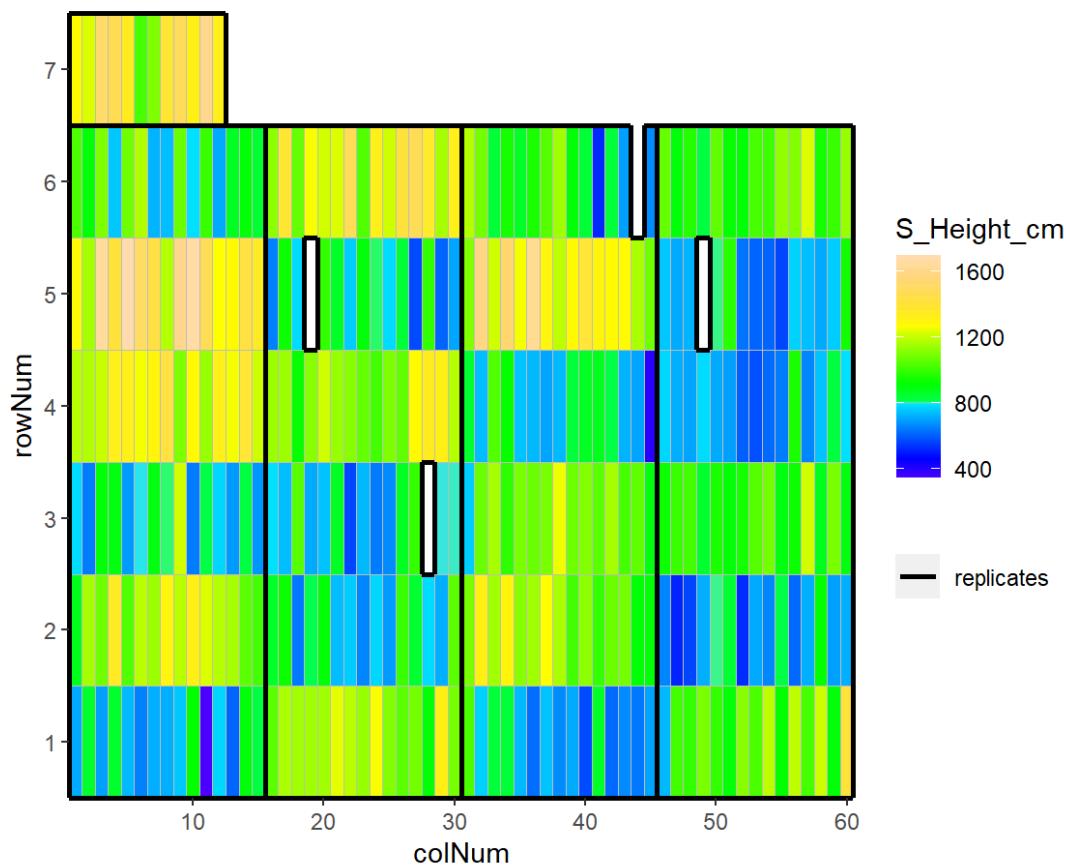
## EPPN2020\_M3P - 2020-03-27



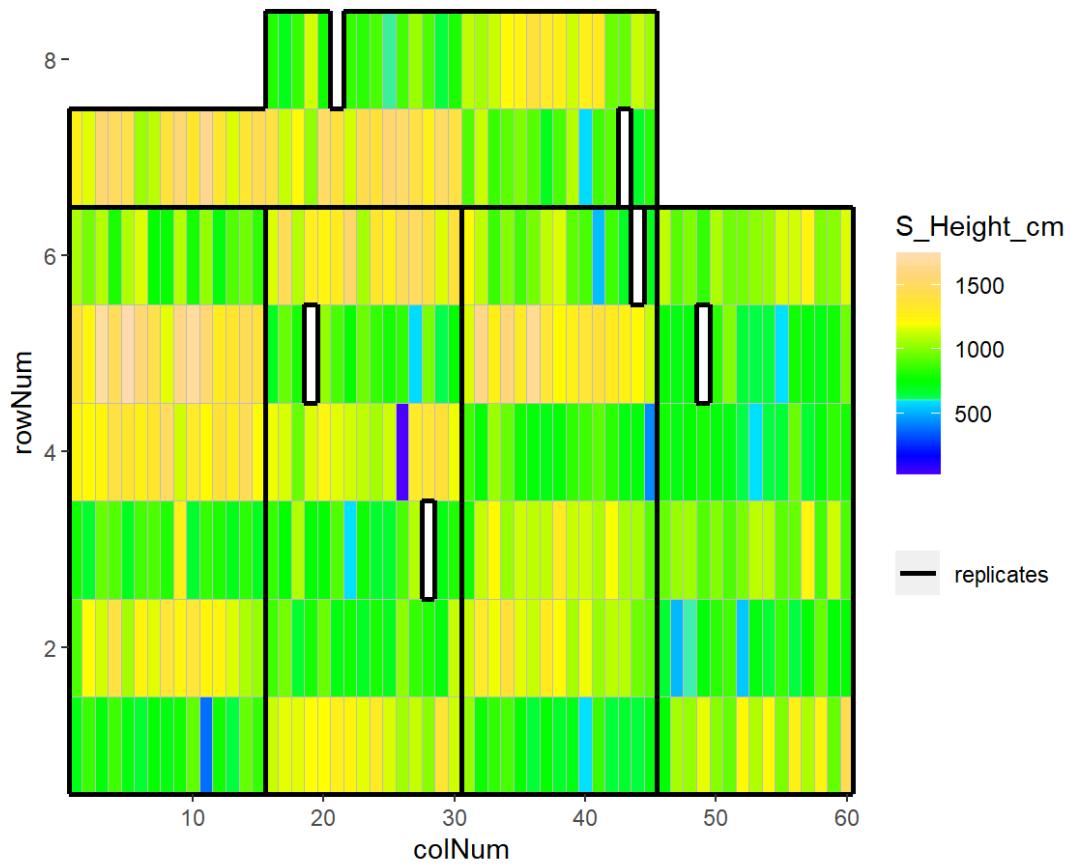
## EPPN2020\_M3P - 2020-03-28



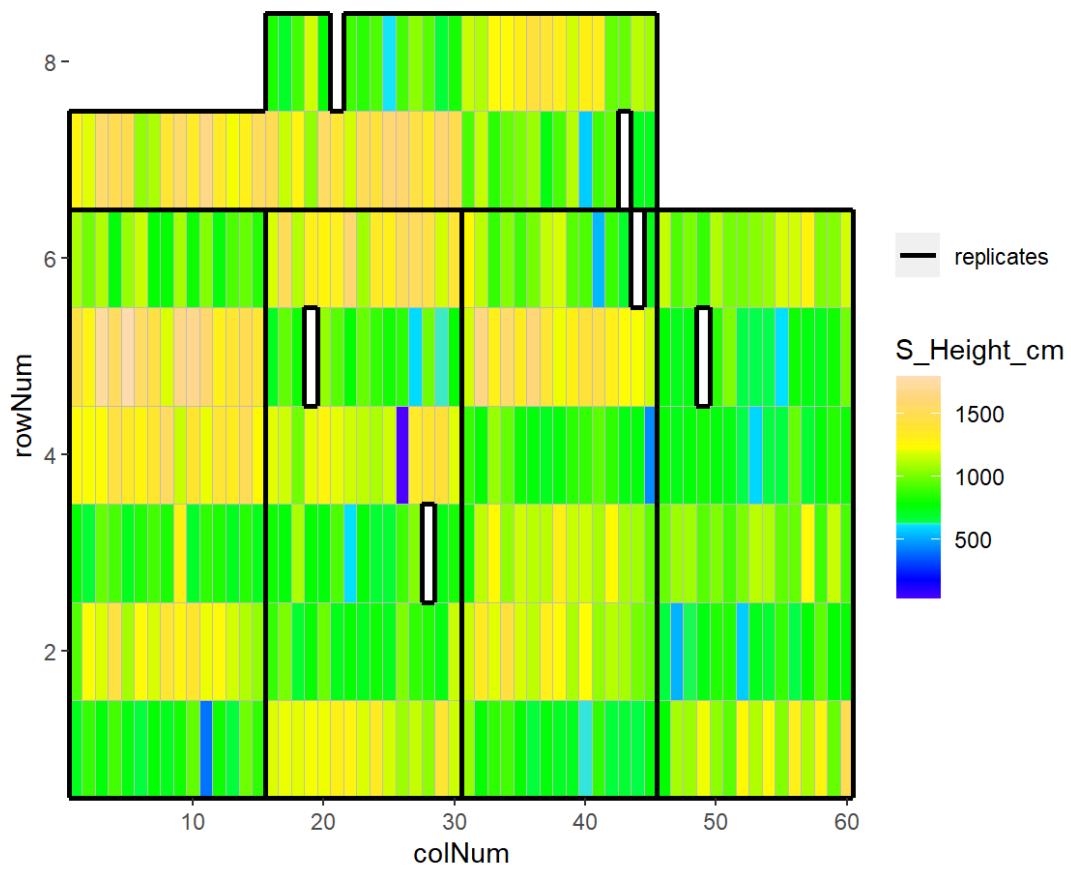
## EPPN2020\_M3P - 2020-03-29



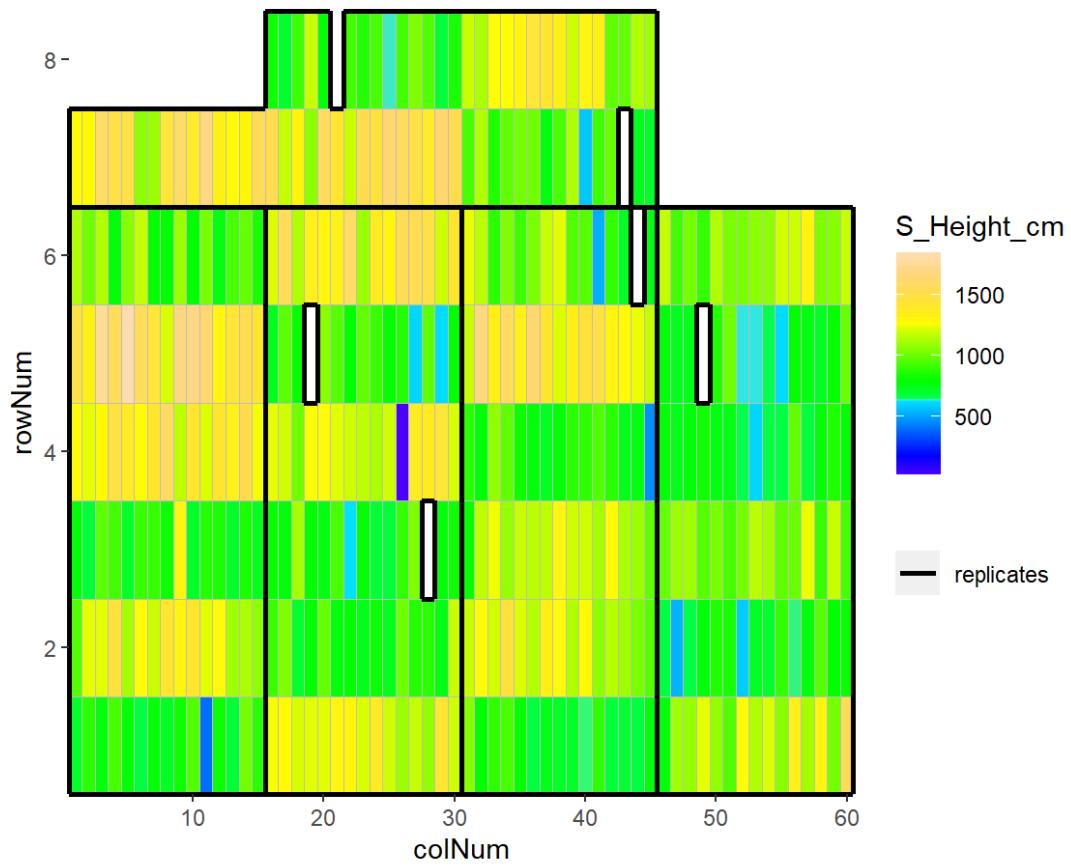
## EPPN2020\_M3P - 2020-03-30



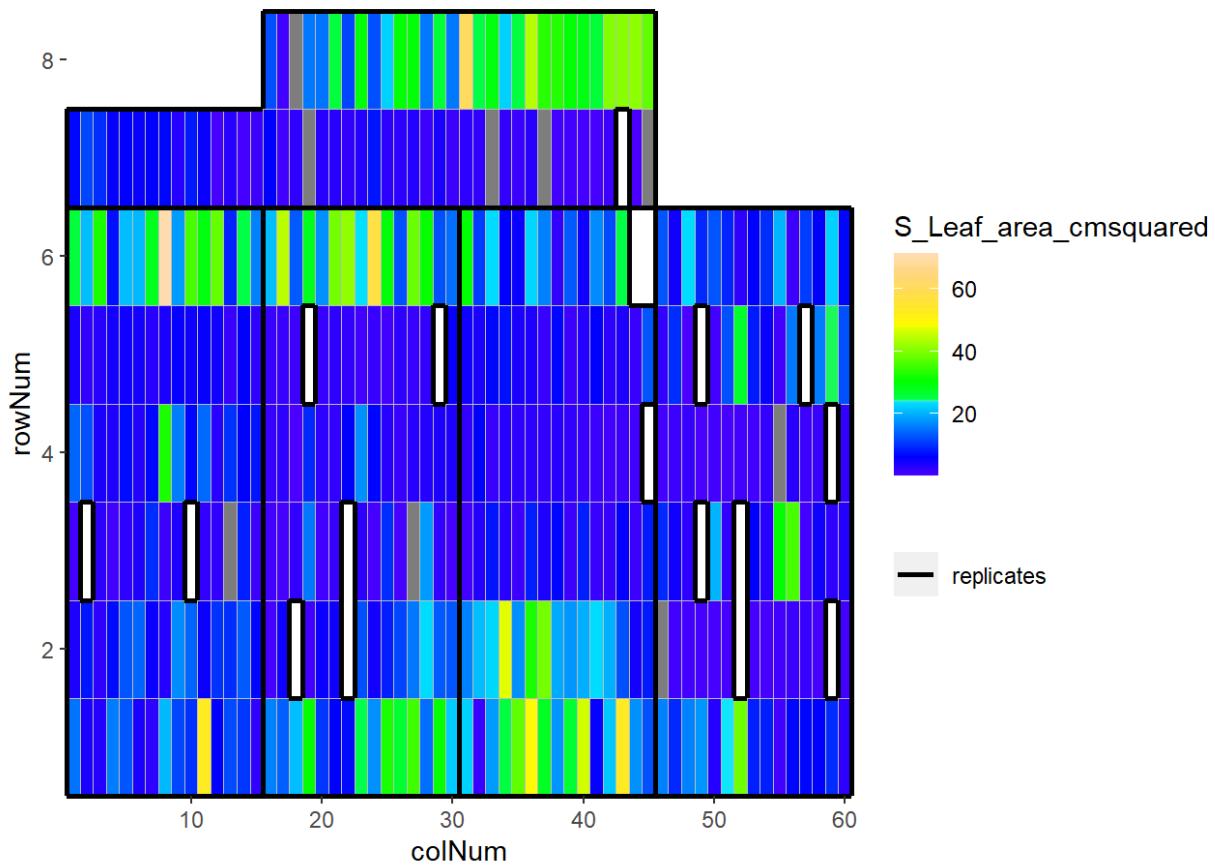
## EPPN2020\_M3P - 2020-03-31



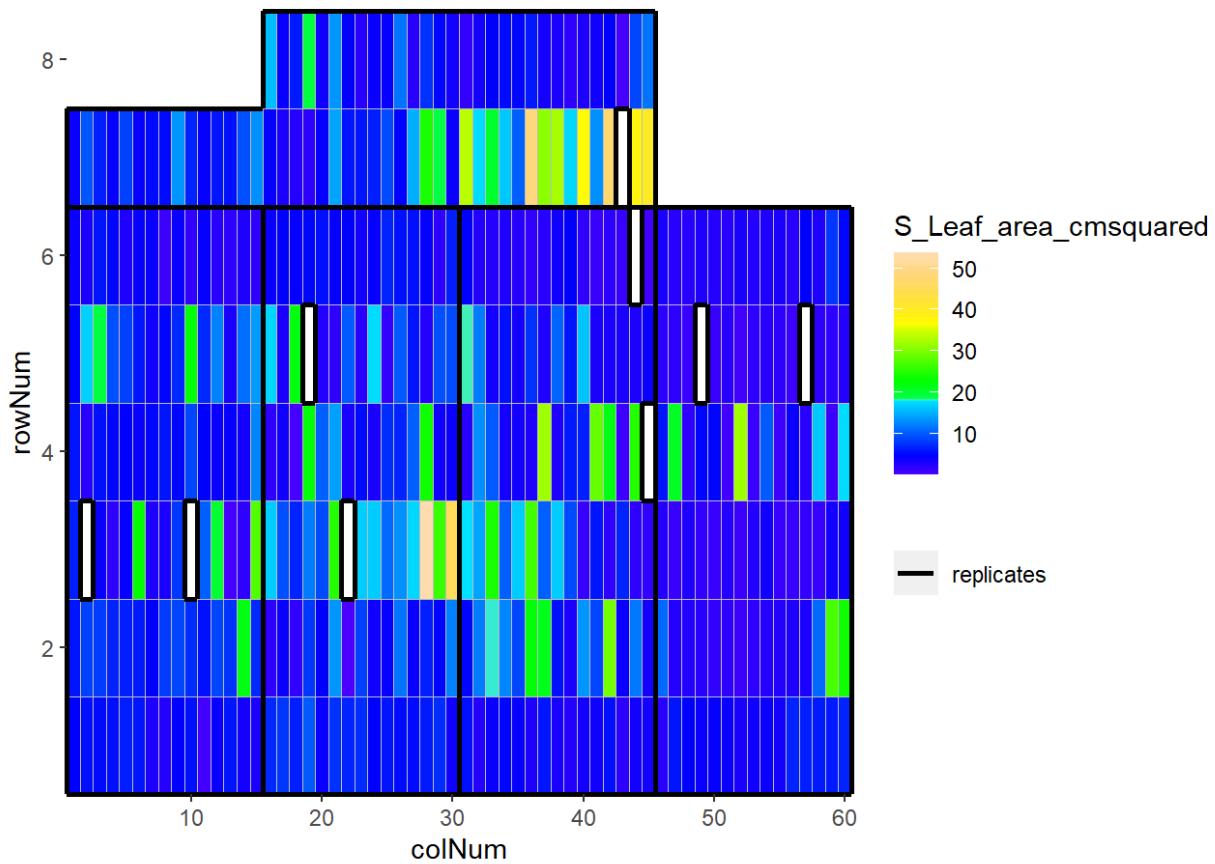
## EPPN2020\_M3P - 2020-04-01



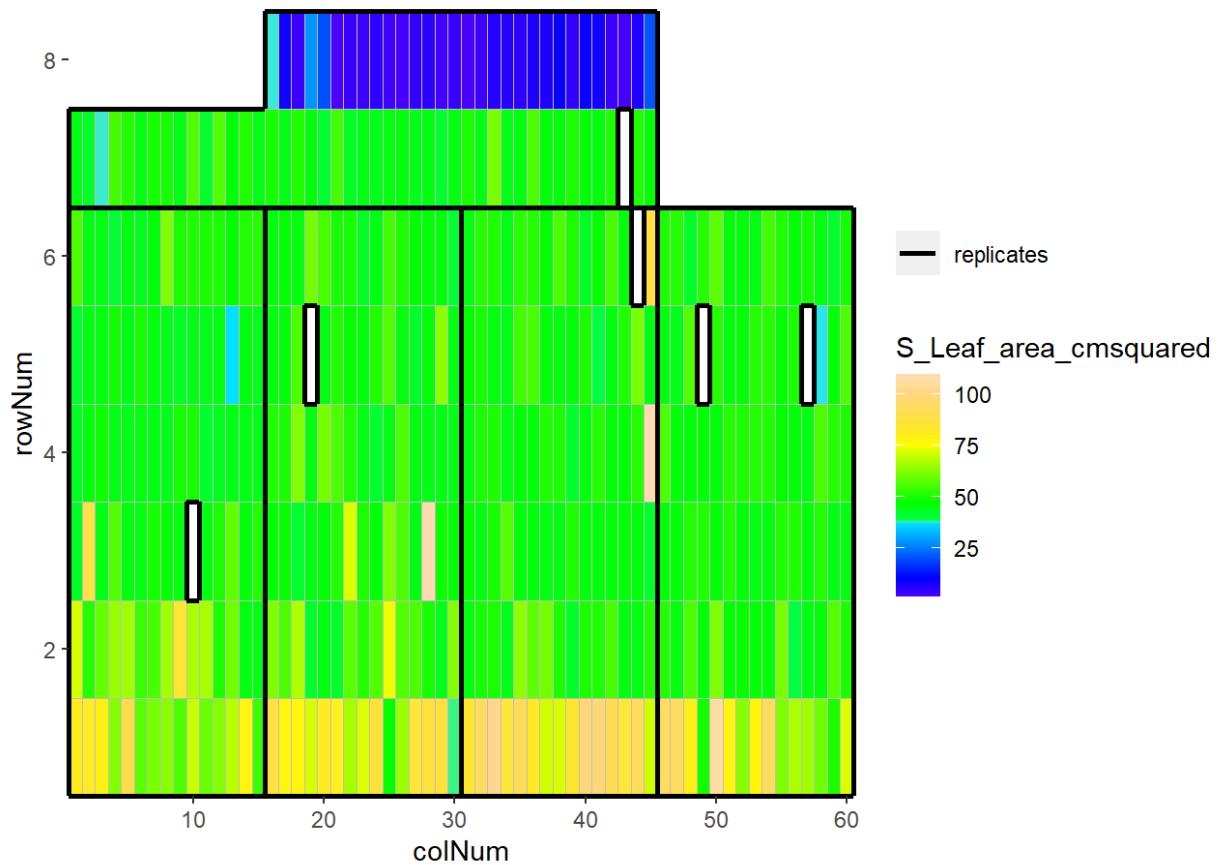
## EPPN2020\_M3P - 2020-02-14



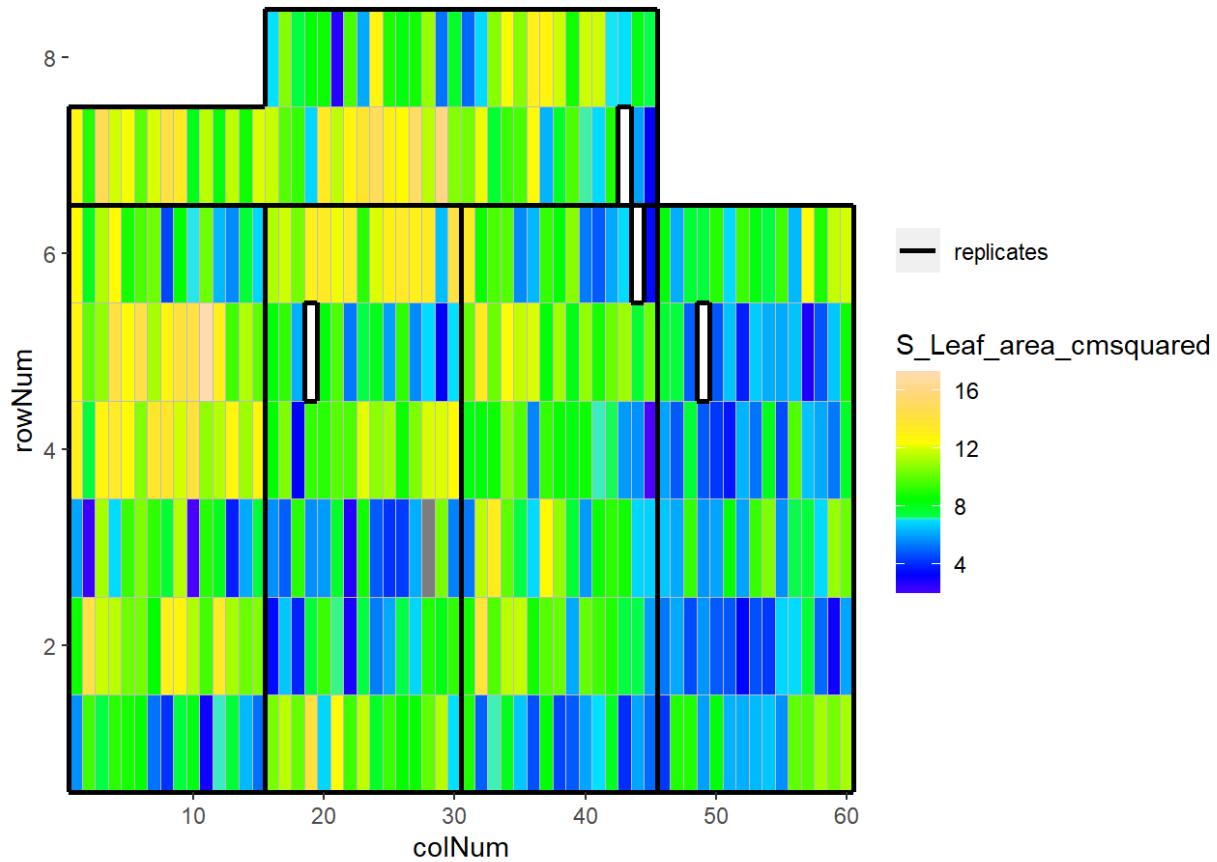
## EPPN2020\_M3P - 2020-02-15



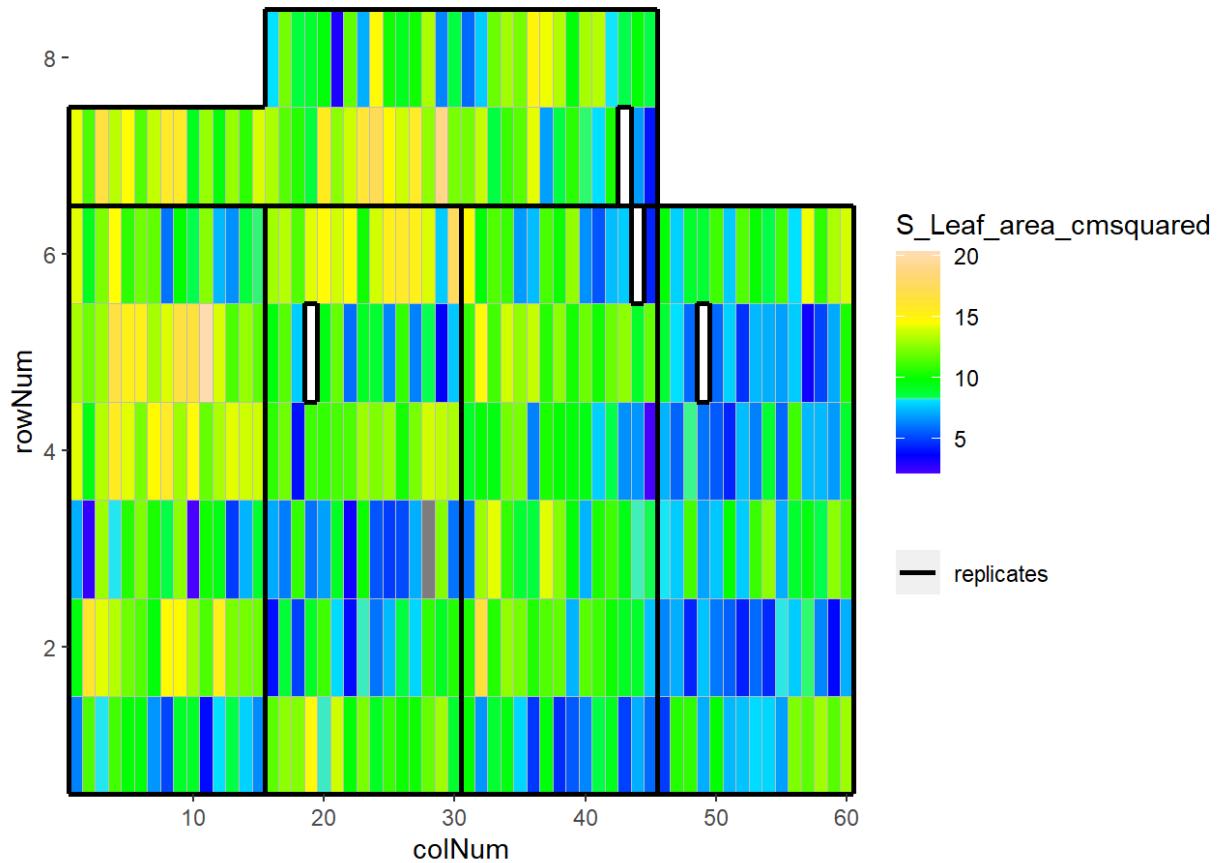
## EPPN2020\_M3P - 2020-02-16



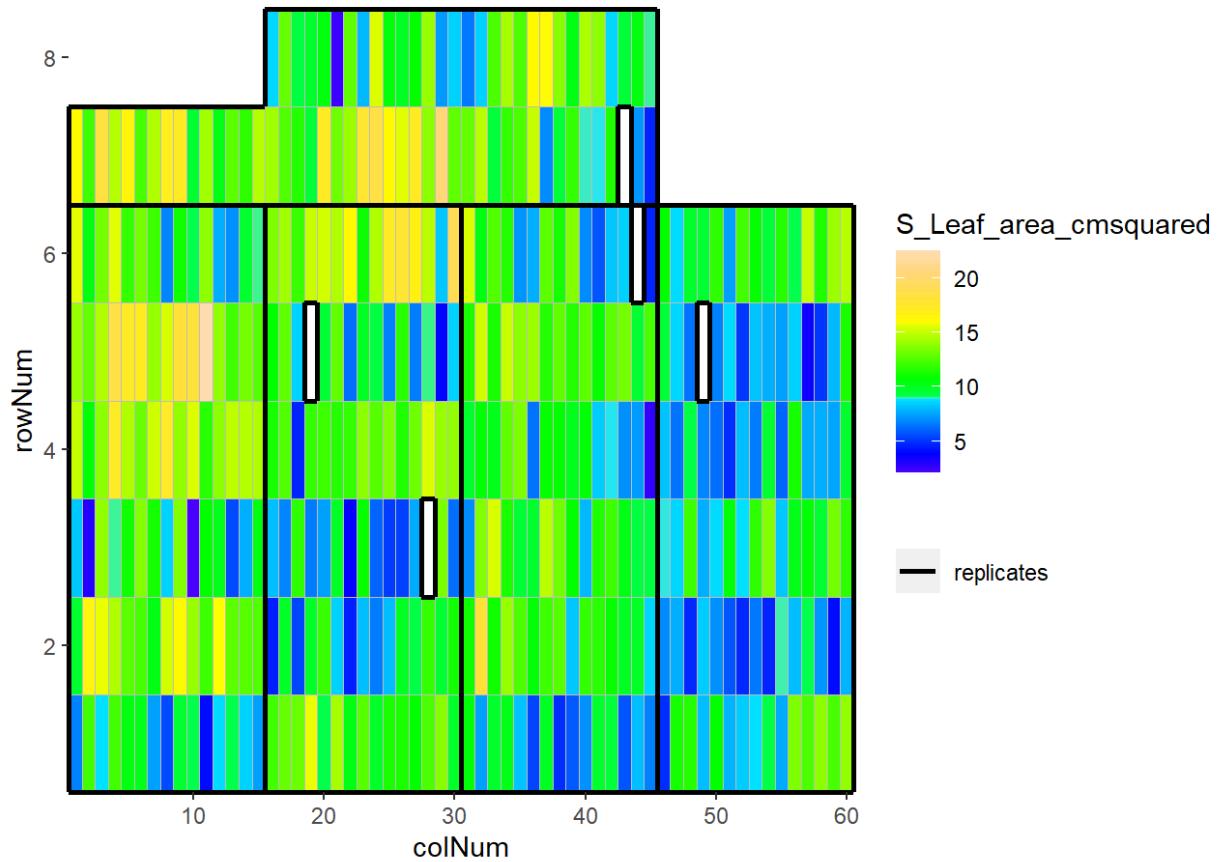
## EPPN2020\_M3P - 2020-02-17



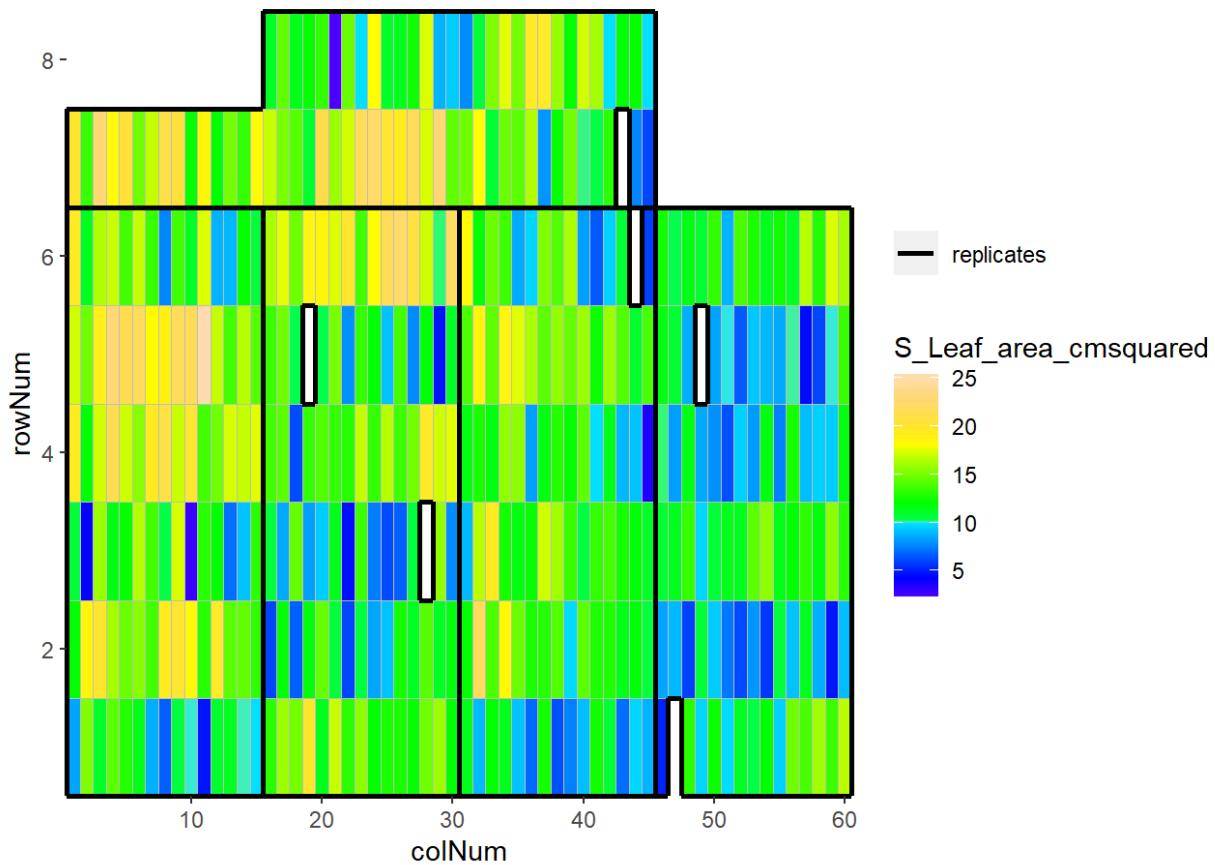
## EPPN2020\_M3P - 2020-02-18



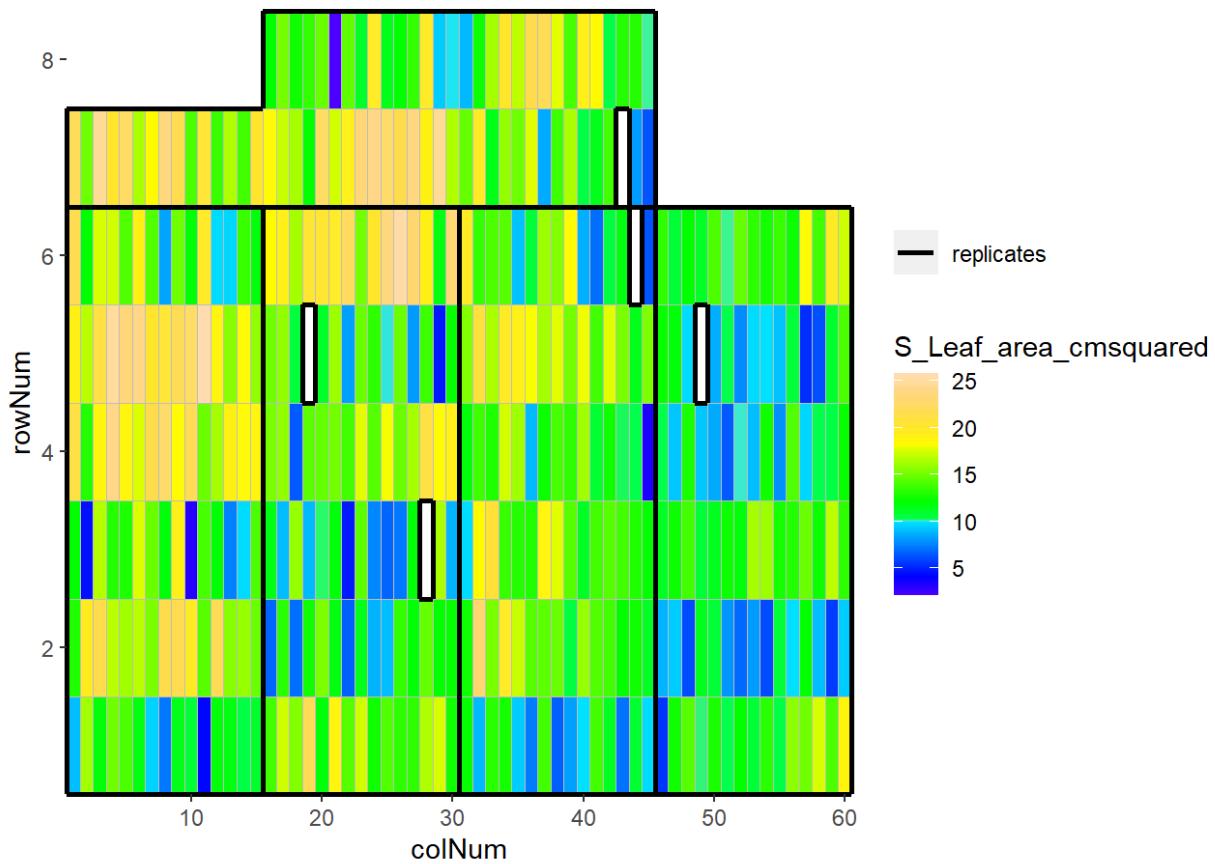
## EPPN2020\_M3P - 2020-02-19



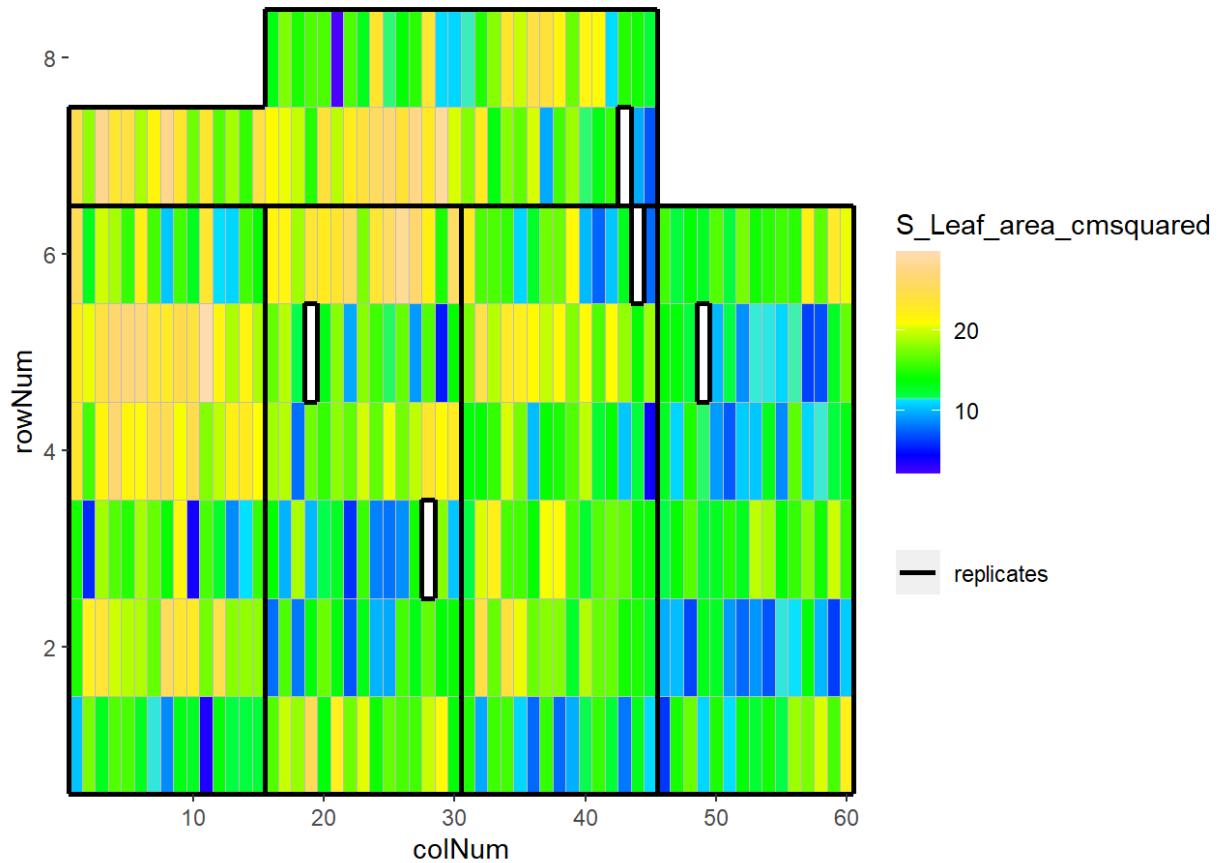
## EPPN2020\_M3P - 2020-02-20



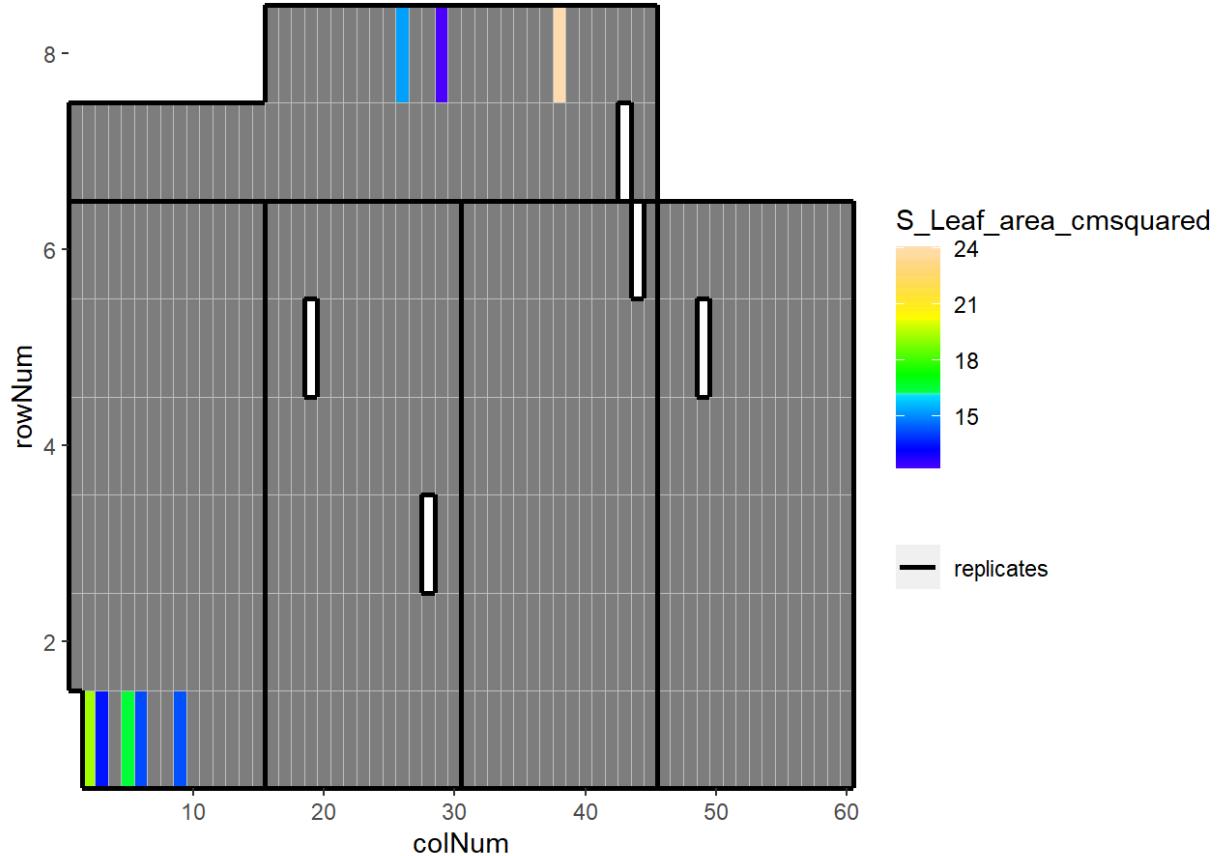
## EPPN2020\_M3P - 2020-02-21



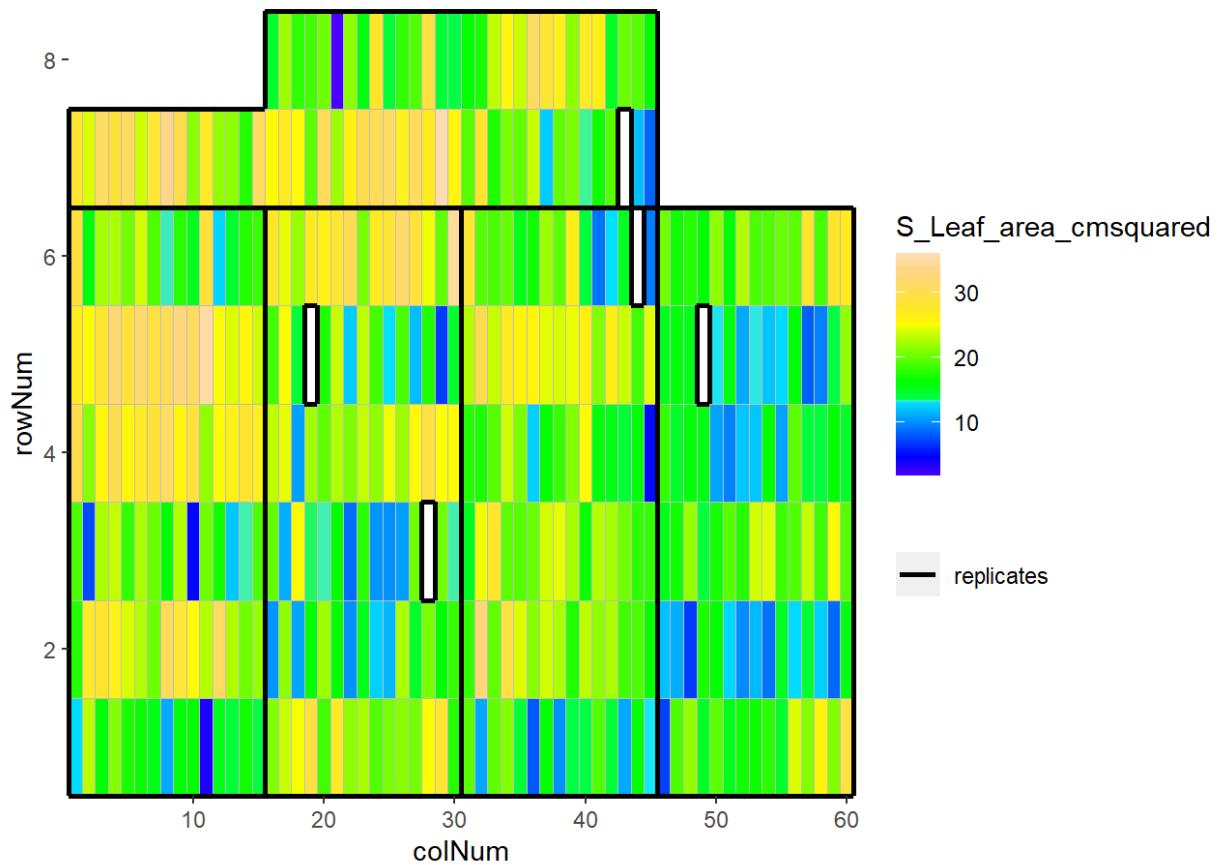
## EPPN2020\_M3P - 2020-02-22



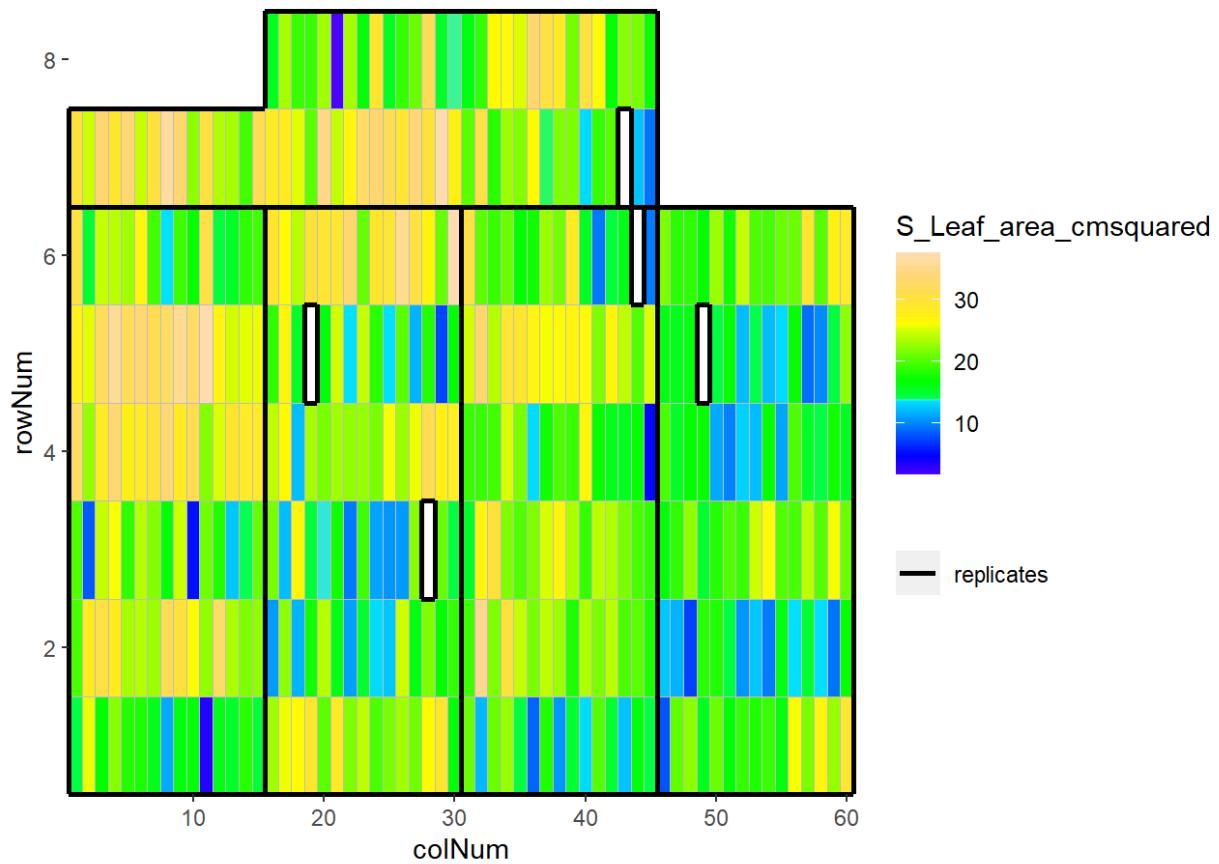
## EPPN2020\_M3P - 2020-02-23



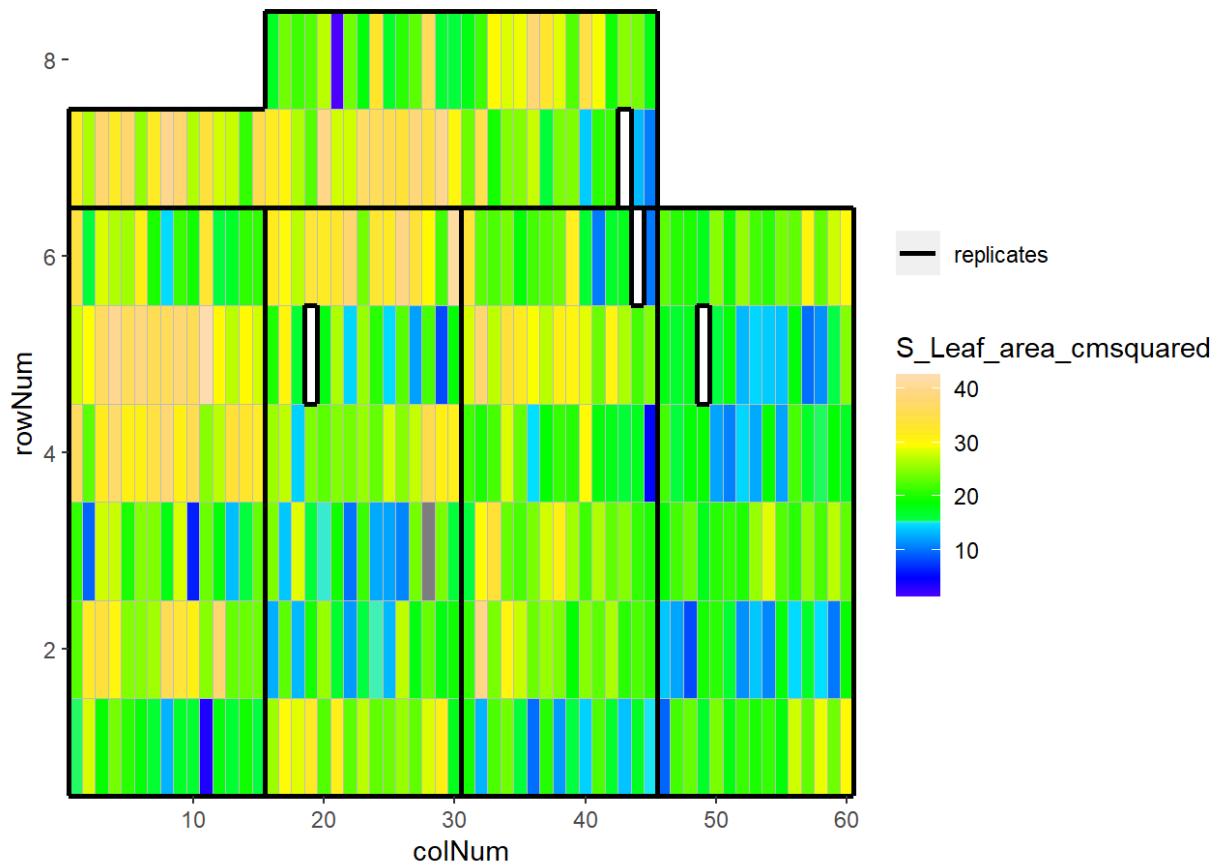
## EPPN2020\_M3P - 2020-02-24



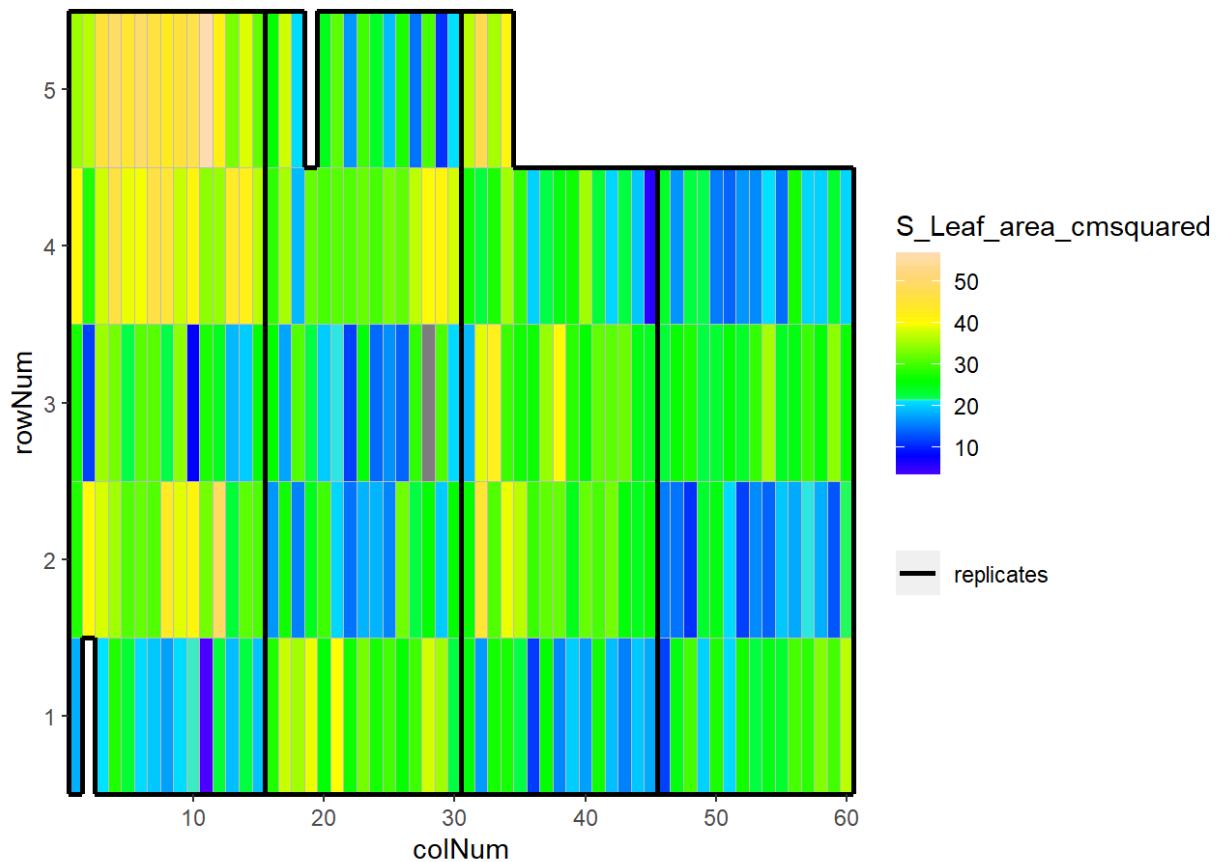
## EPPN2020\_M3P - 2020-02-25



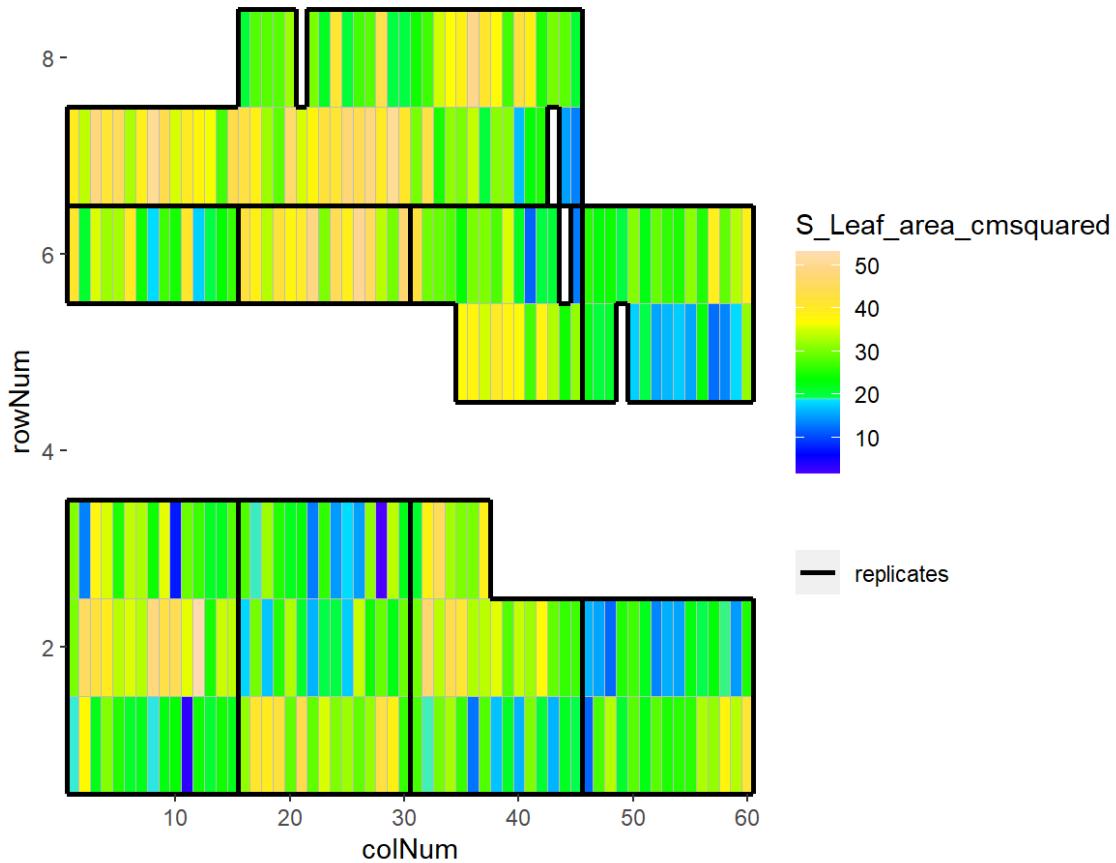
## EPPN2020\_M3P - 2020-02-26



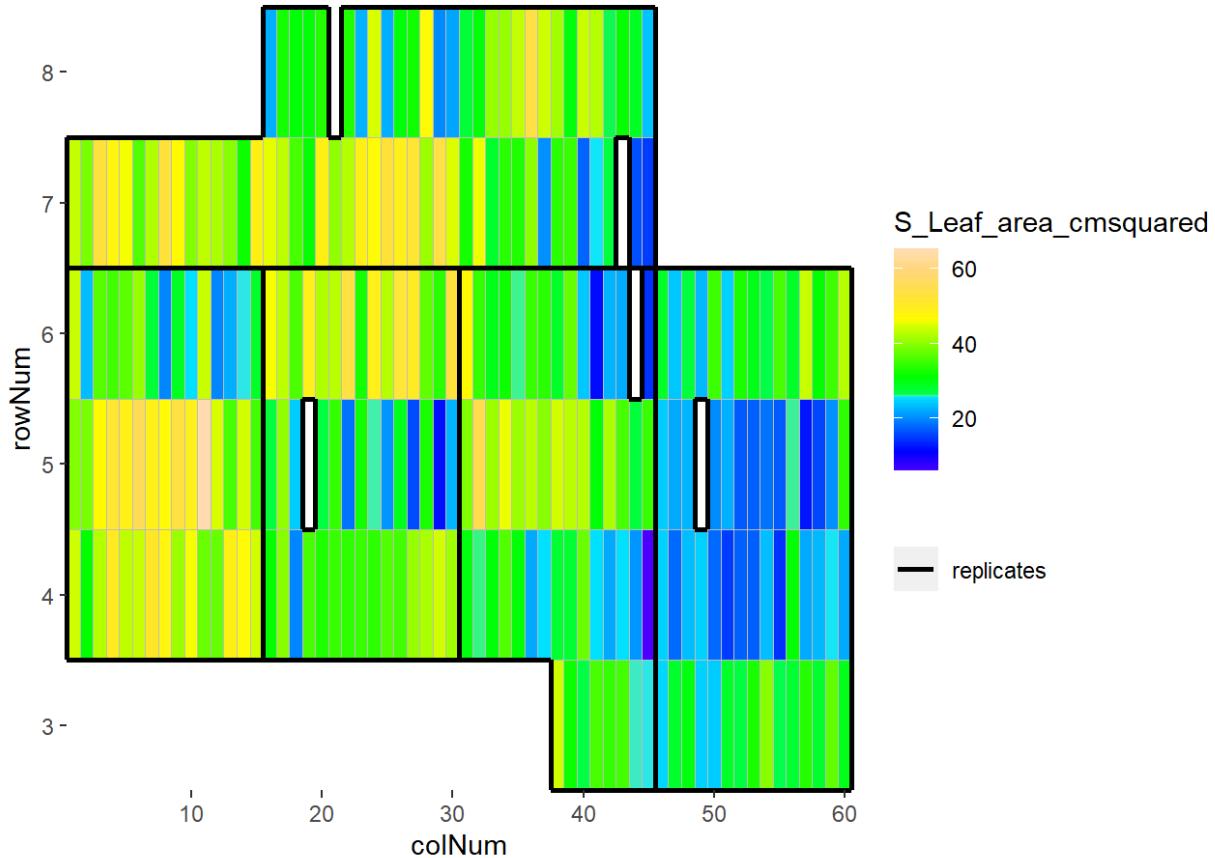
## EPPN2020\_M3P - 2020-02-28



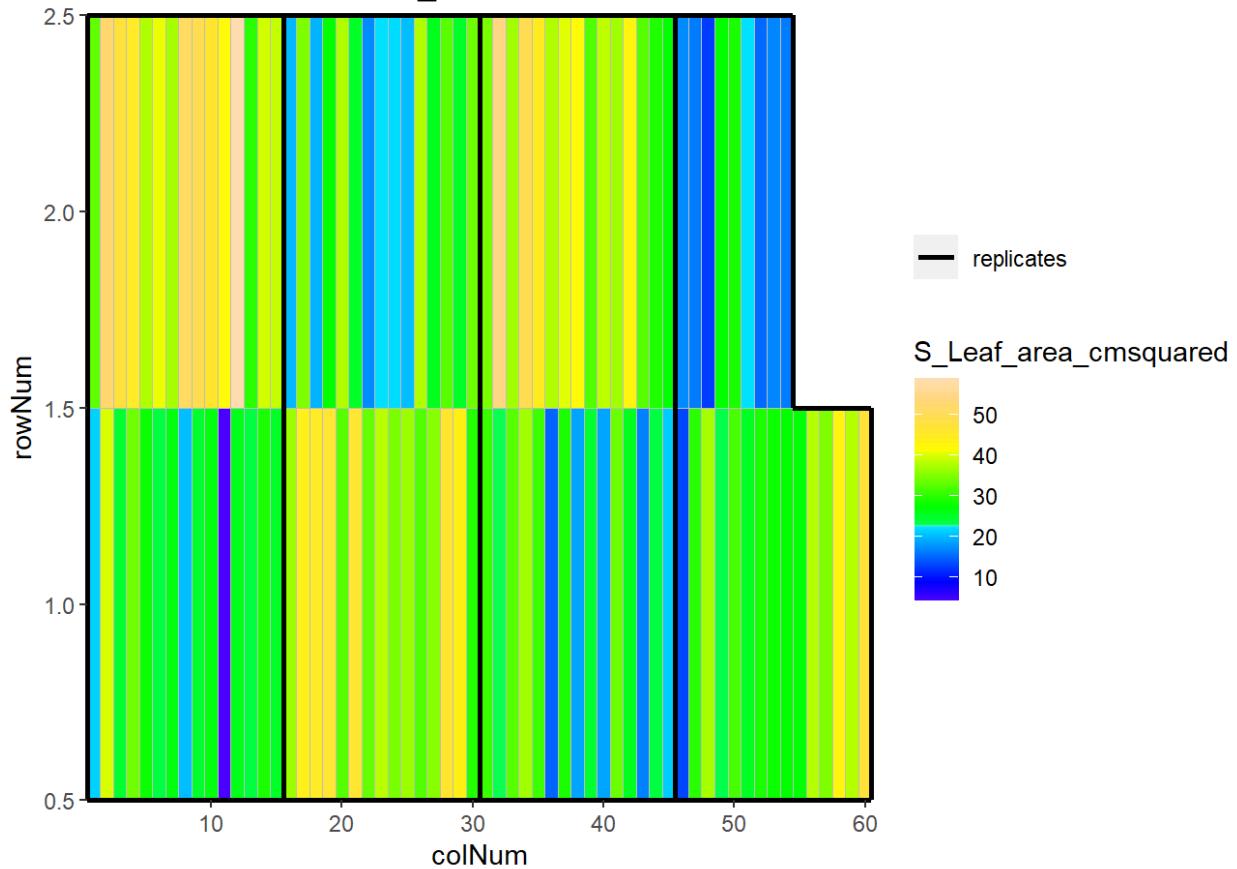
## EPPN2020\_M3P - 2020-02-29



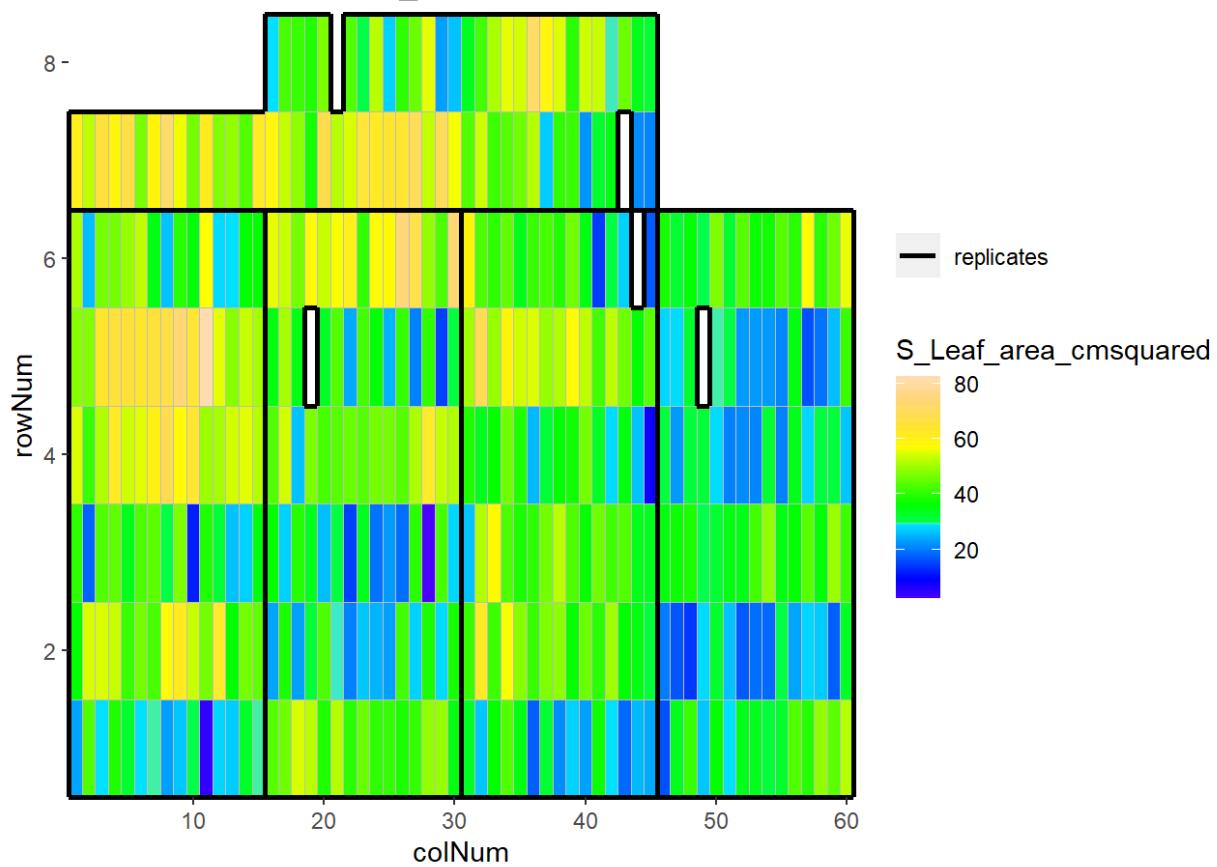
## EPPN2020\_M3P - 2020-03-01



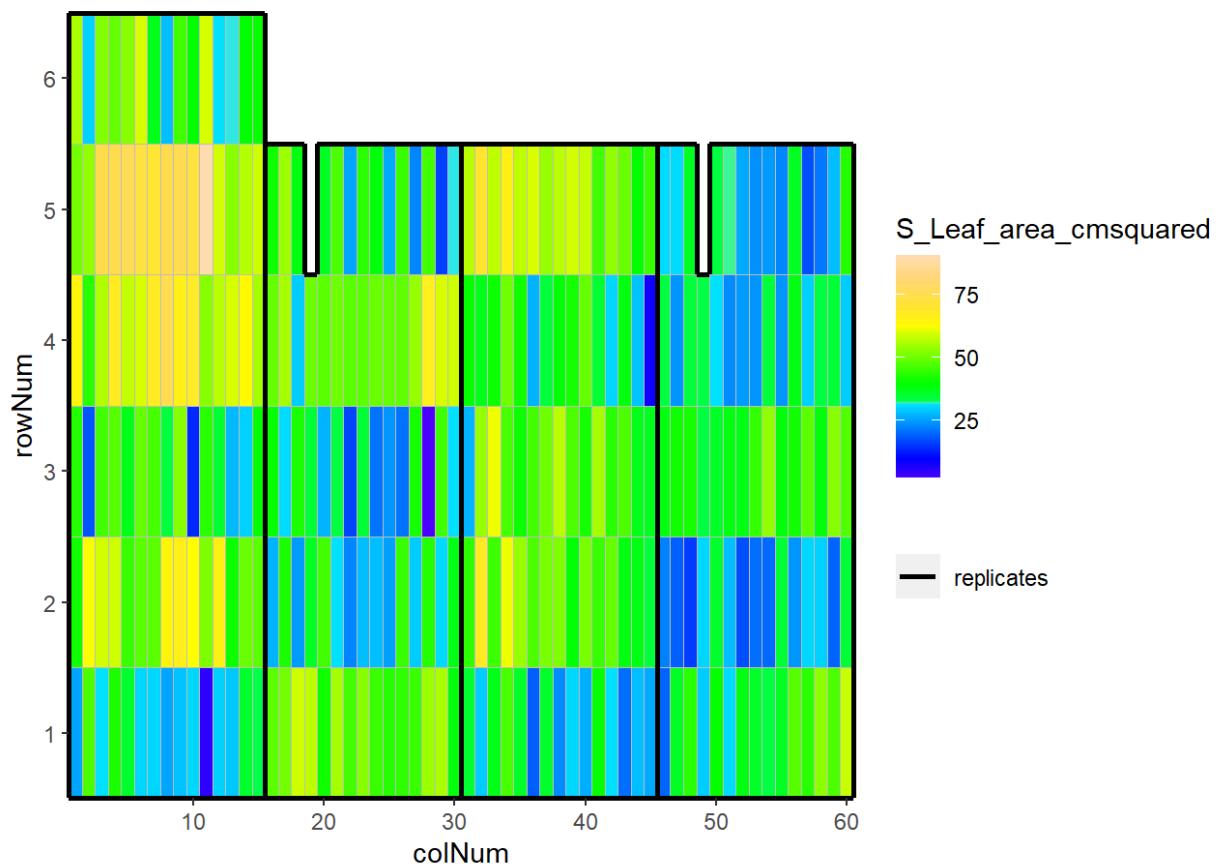
## EPPN2020\_M3P - 2020-03-02



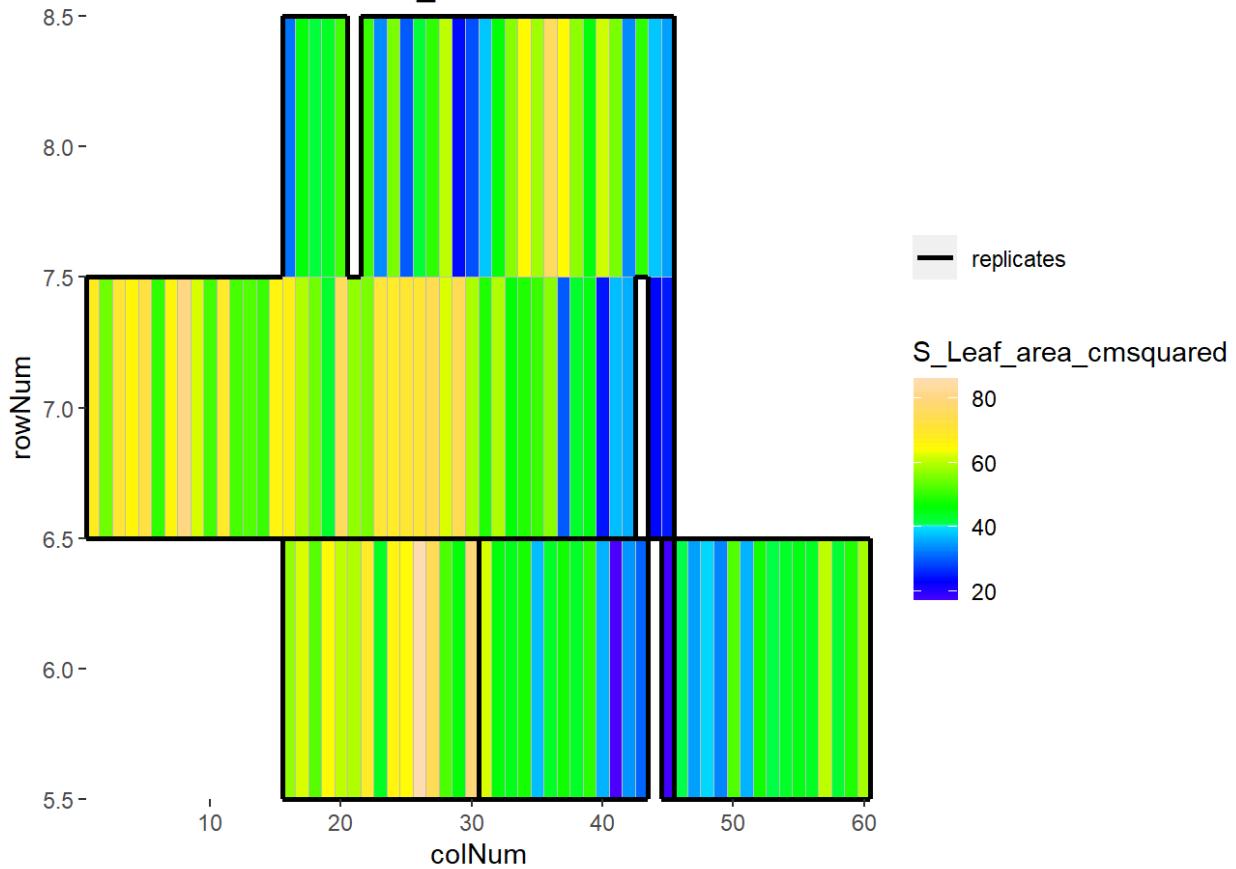
## EPPN2020\_M3P - 2020-03-03



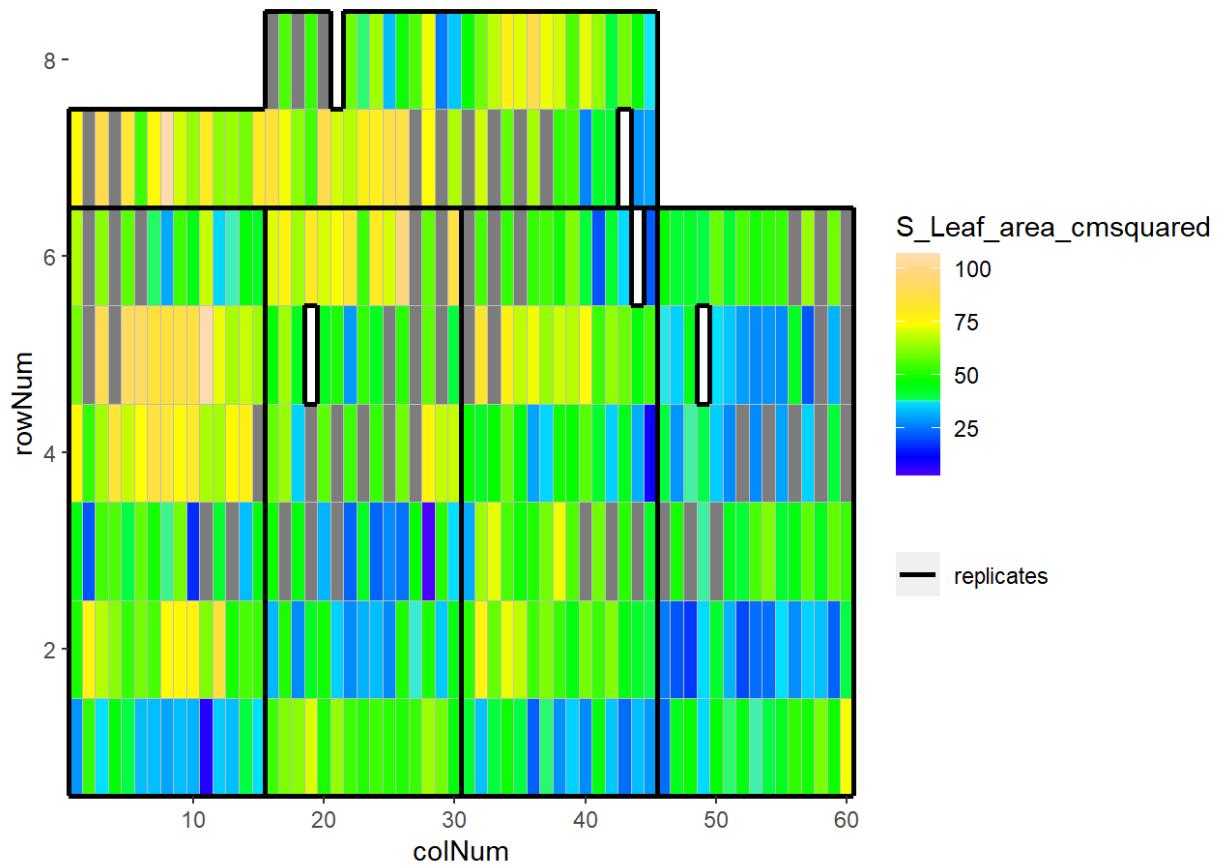
## EPPN2020\_M3P - 2020-03-04



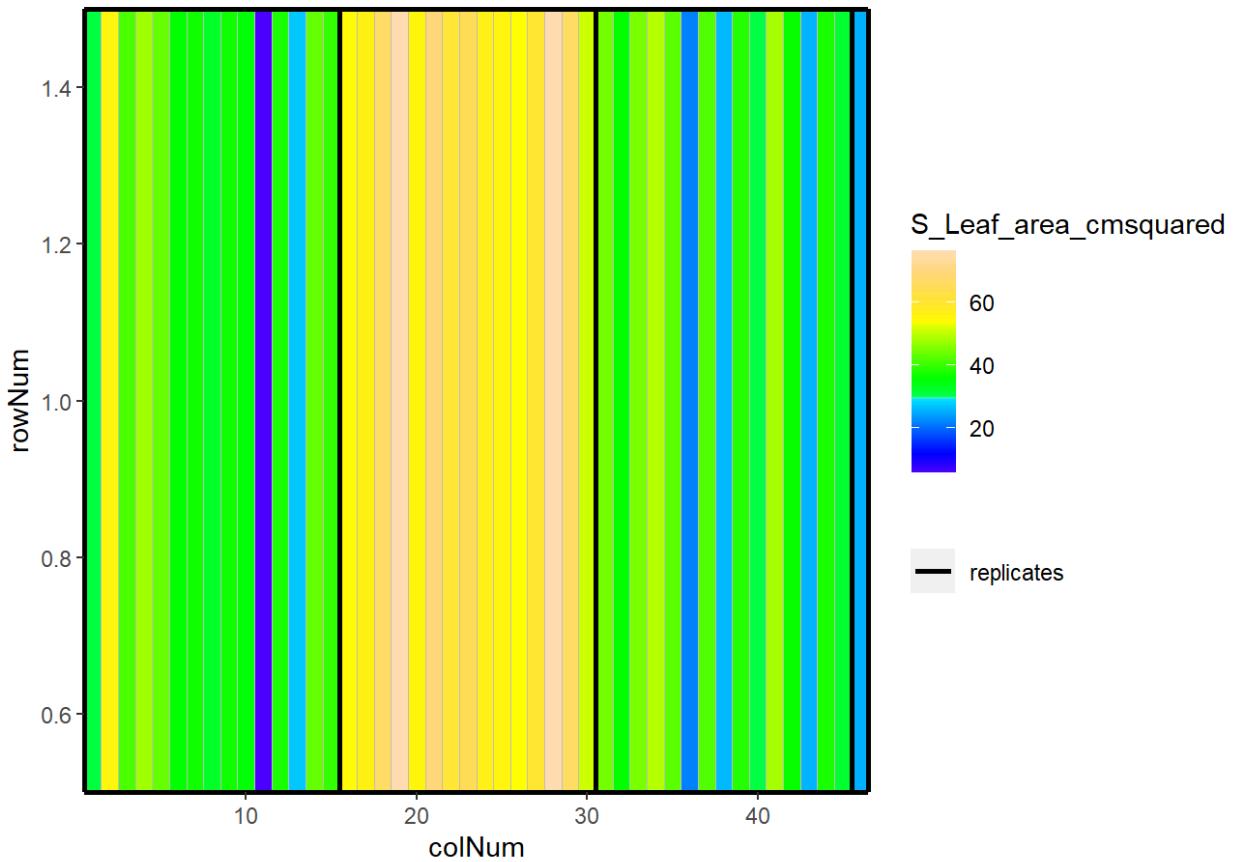
## EPPN2020\_M3P - 2020-03-05



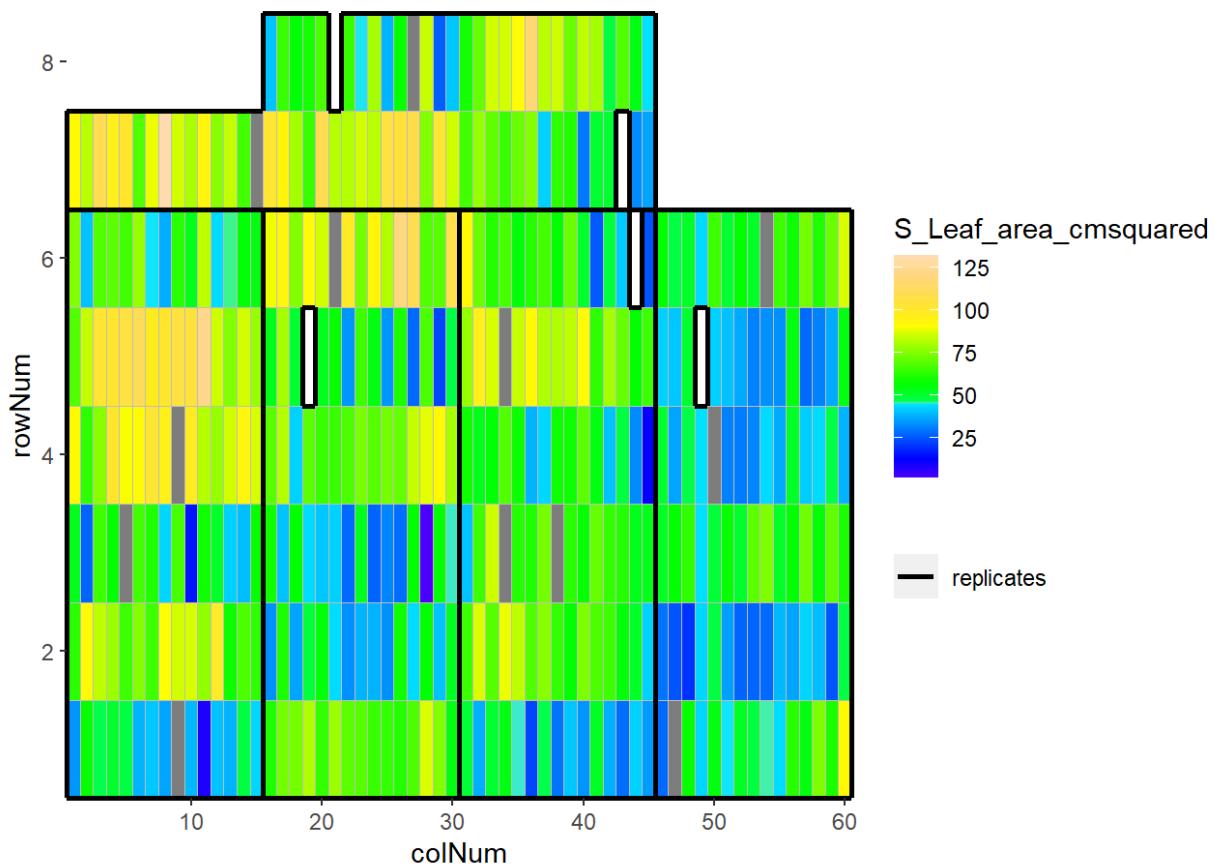
## EPPN2020\_M3P - 2020-03-06



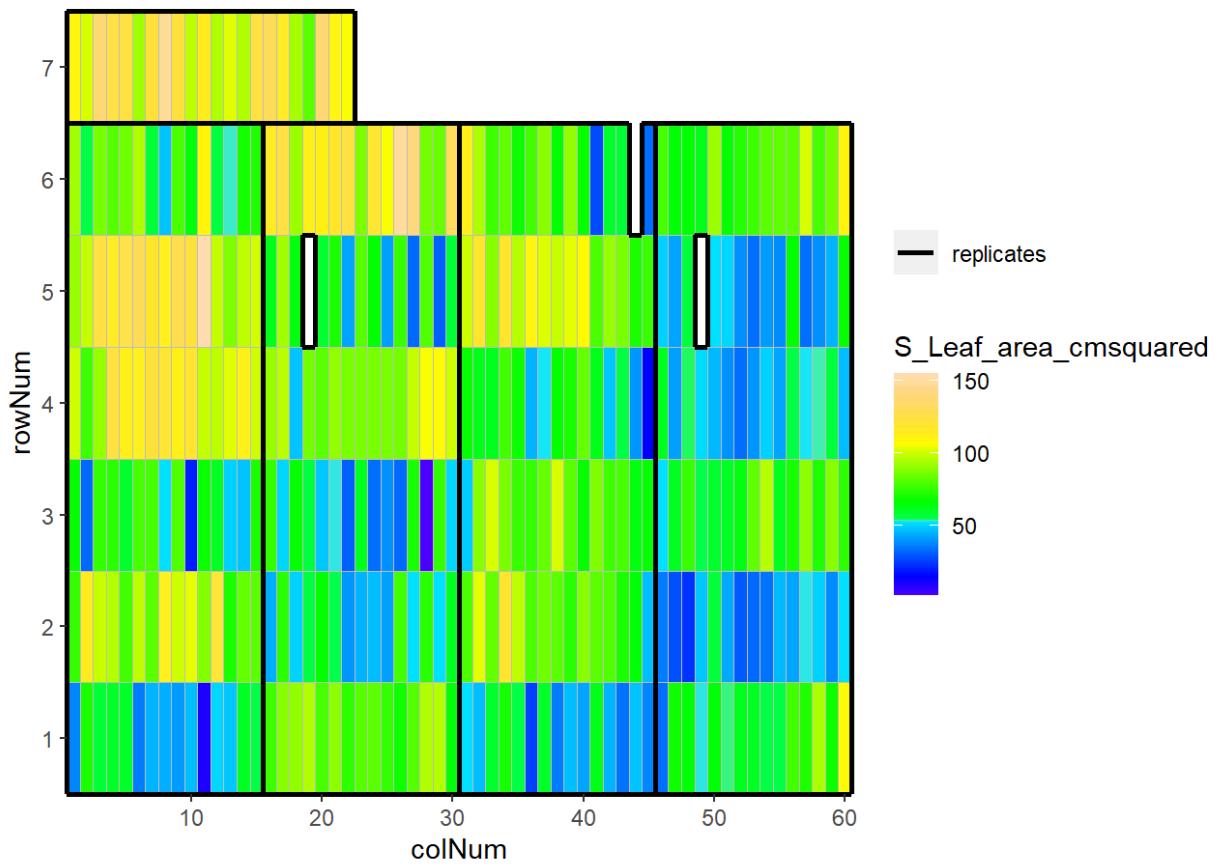
## EPPN2020\_M3P - 2020-03-07



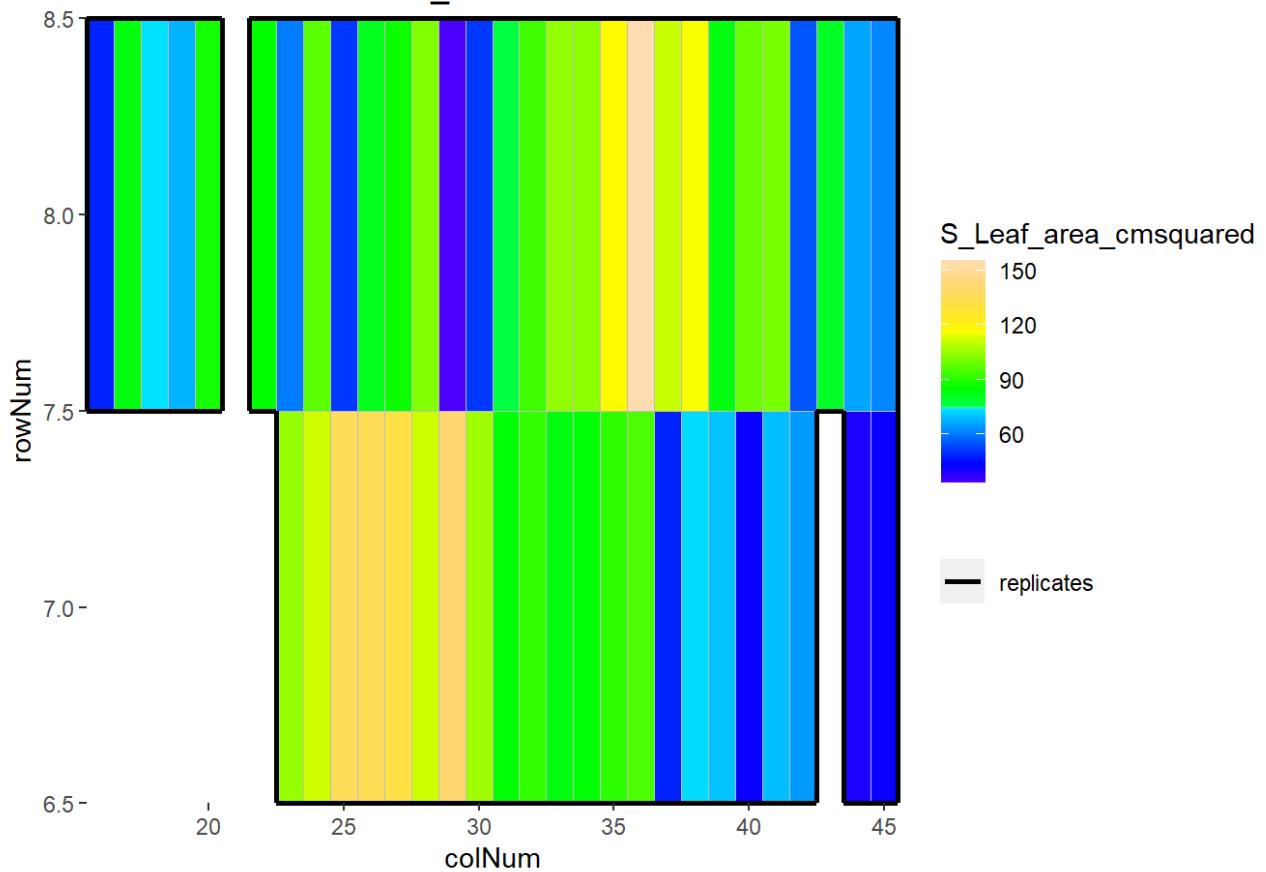
## EPPN2020\_M3P - 2020-03-08



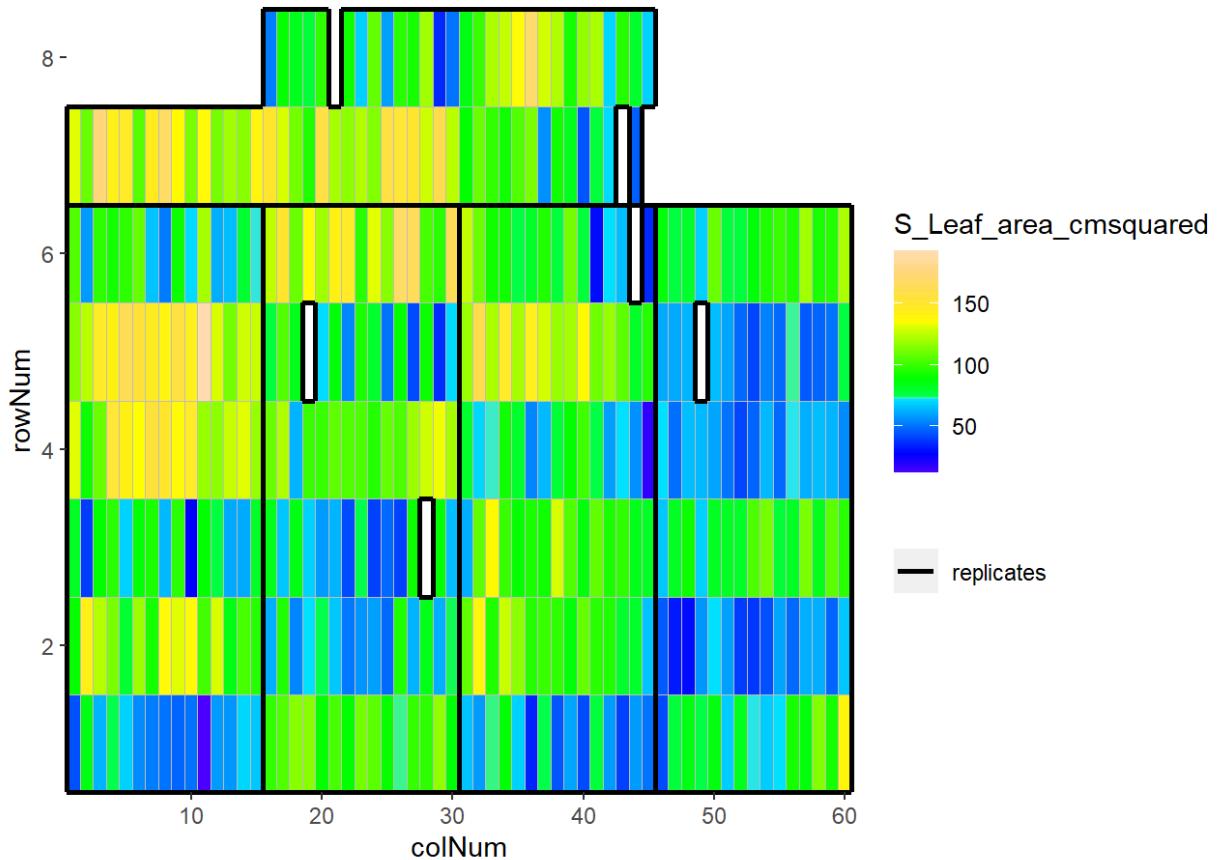
## EPPN2020\_M3P - 2020-03-11



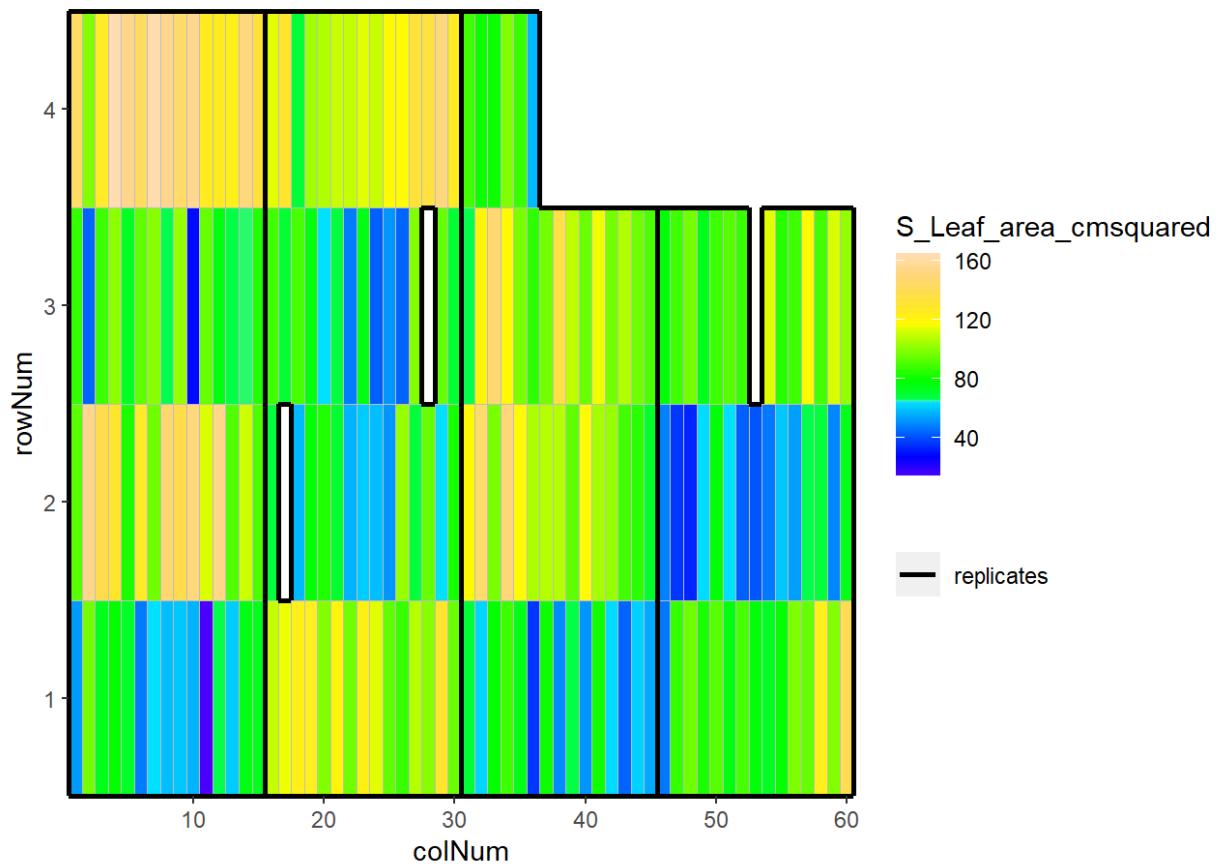
## EPPN2020\_M3P - 2020-03-12



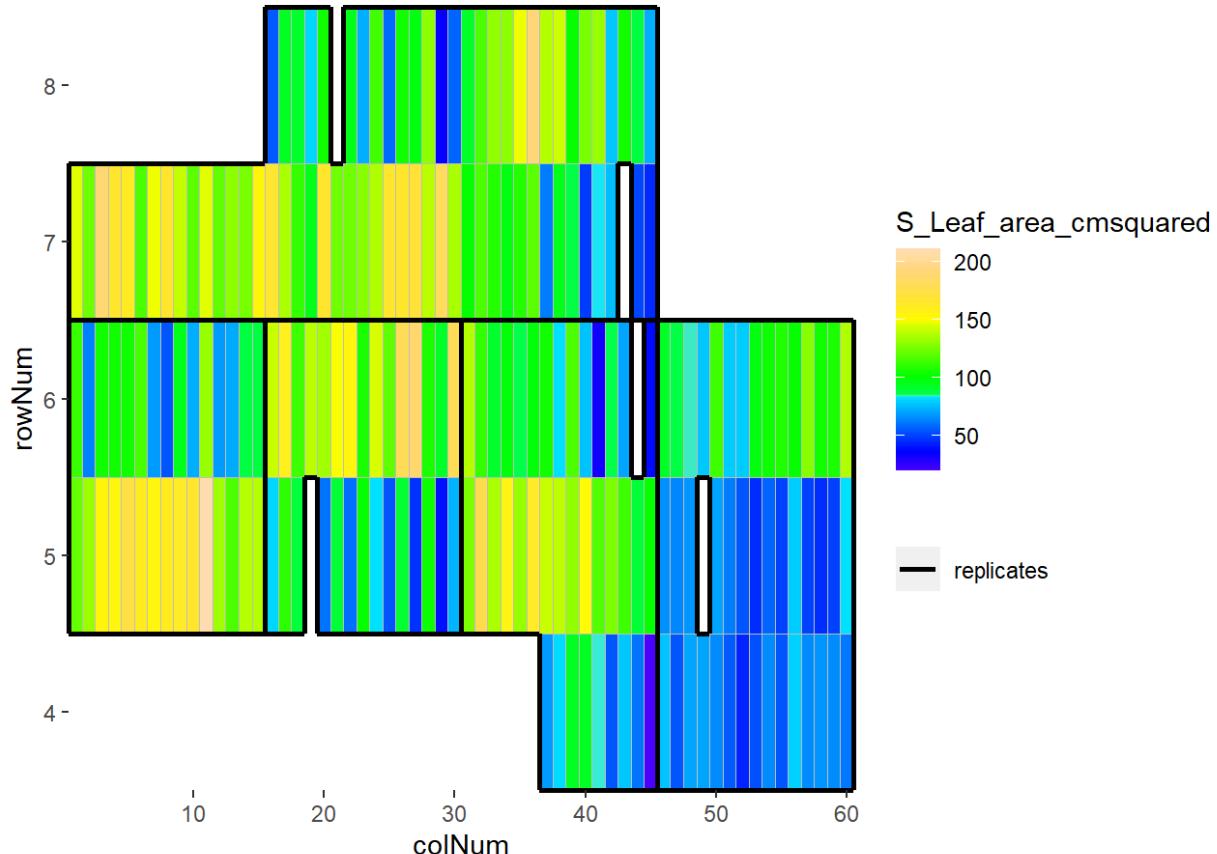
## EPPN2020\_M3P - 2020-03-13



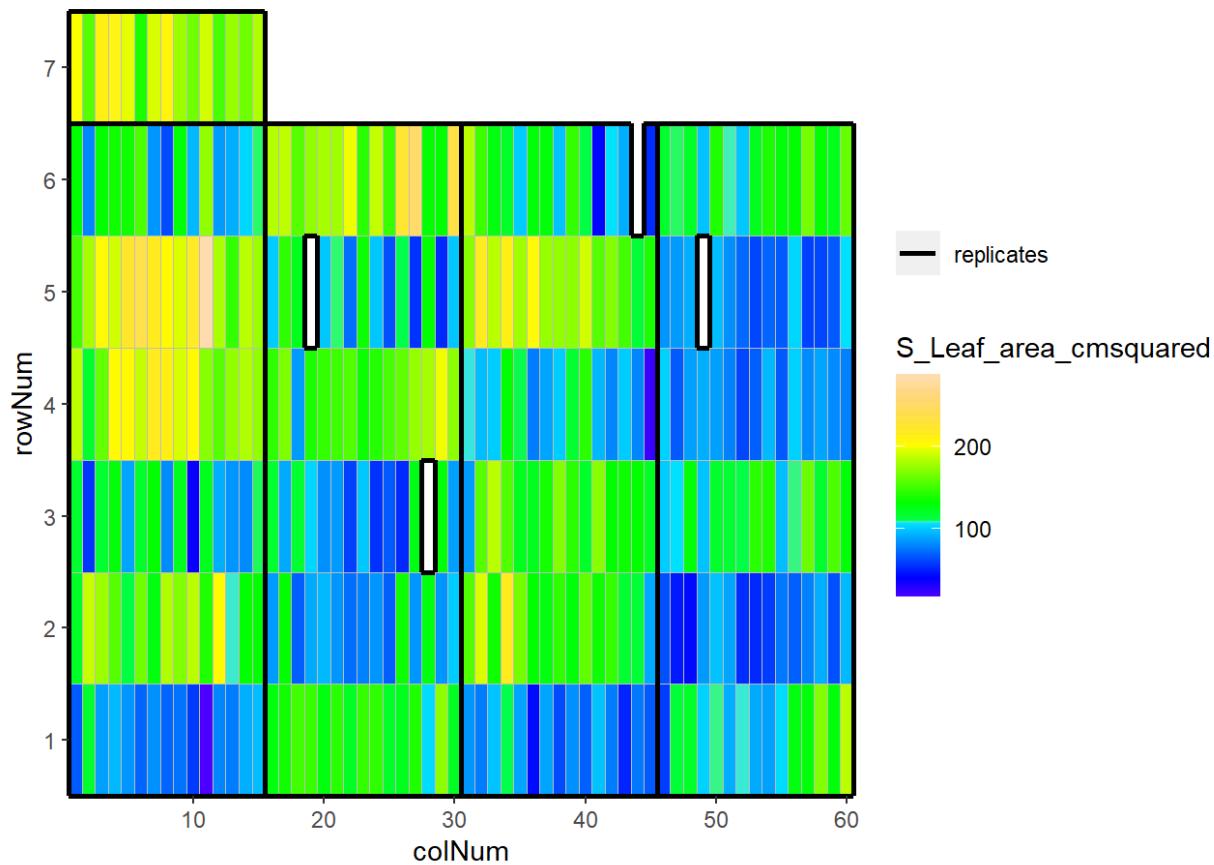
## EPPN2020\_M3P - 2020-03-14



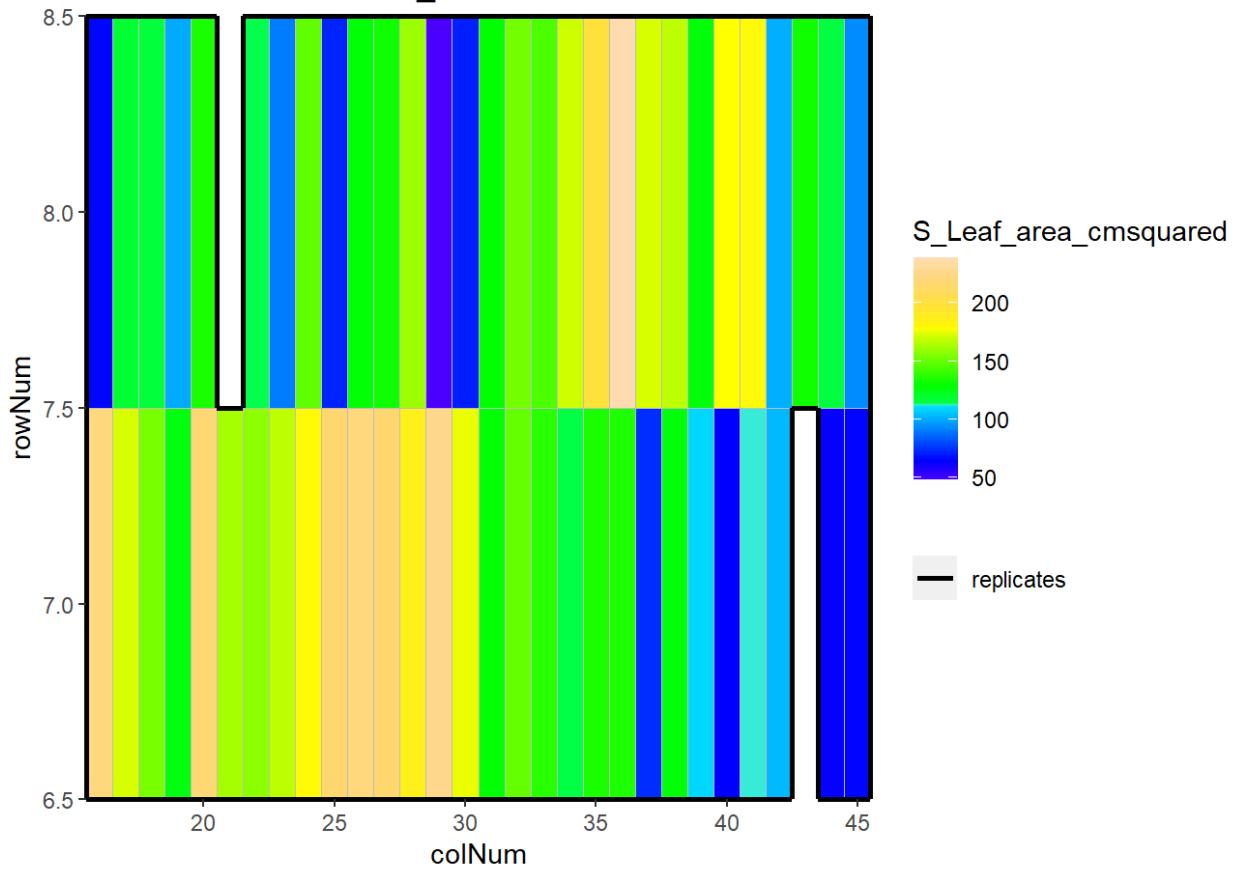
## EPPN2020\_M3P - 2020-03-15



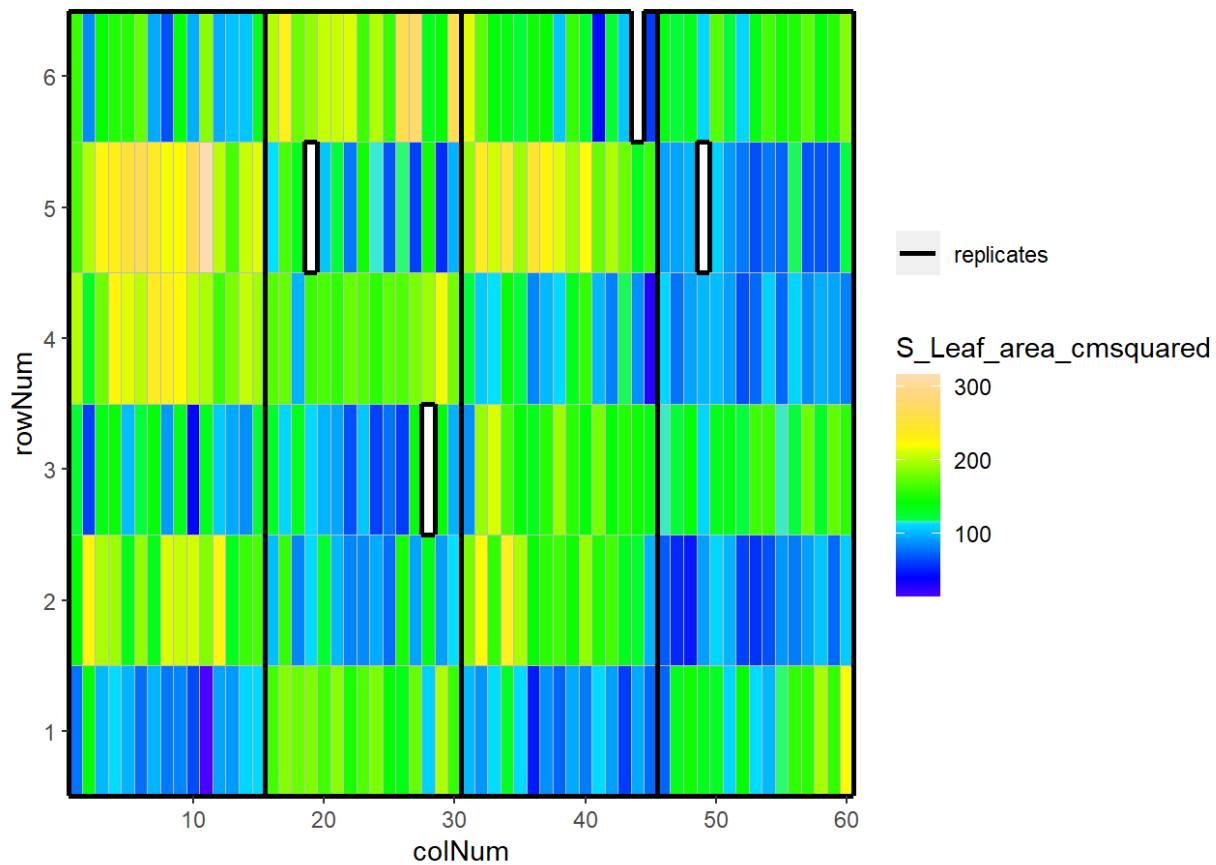
## EPPN2020\_M3P - 2020-03-18



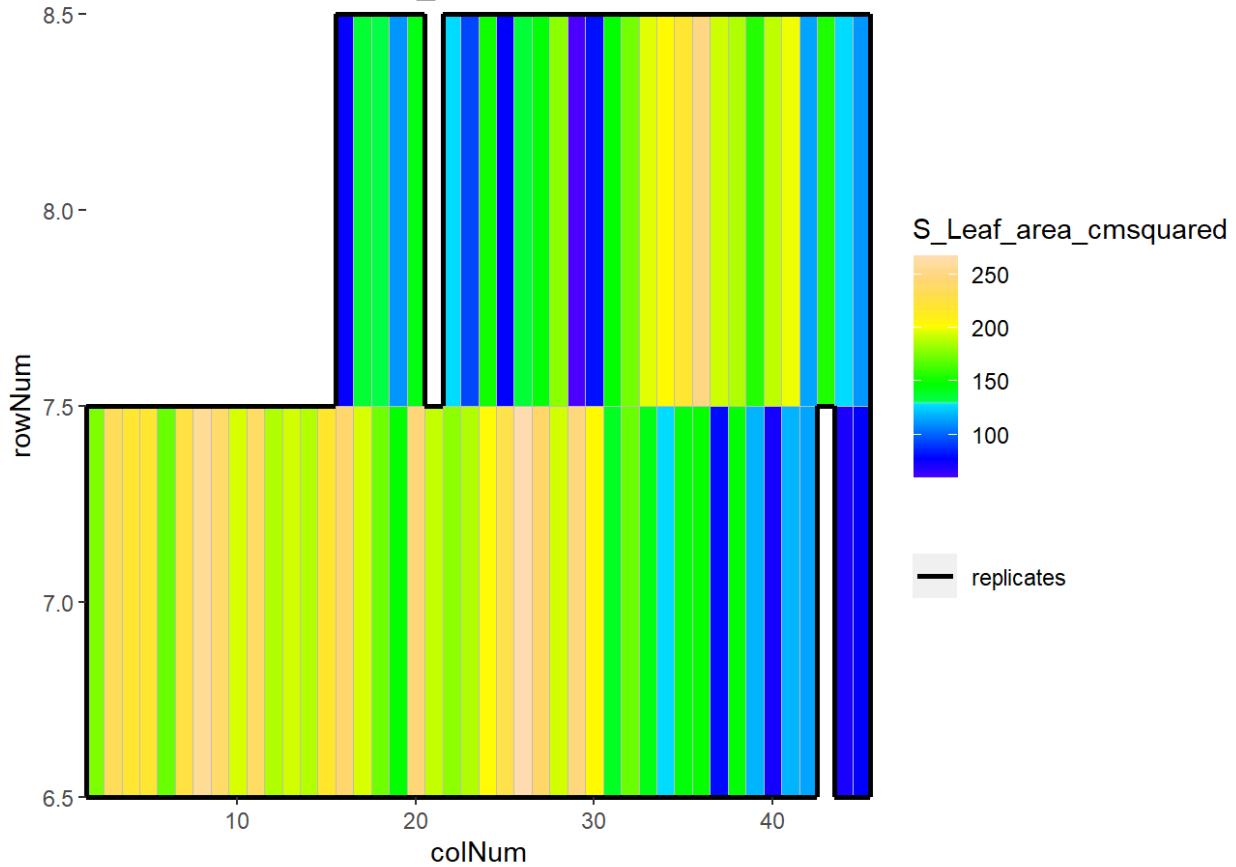
## EPPN2020\_M3P - 2020-03-19



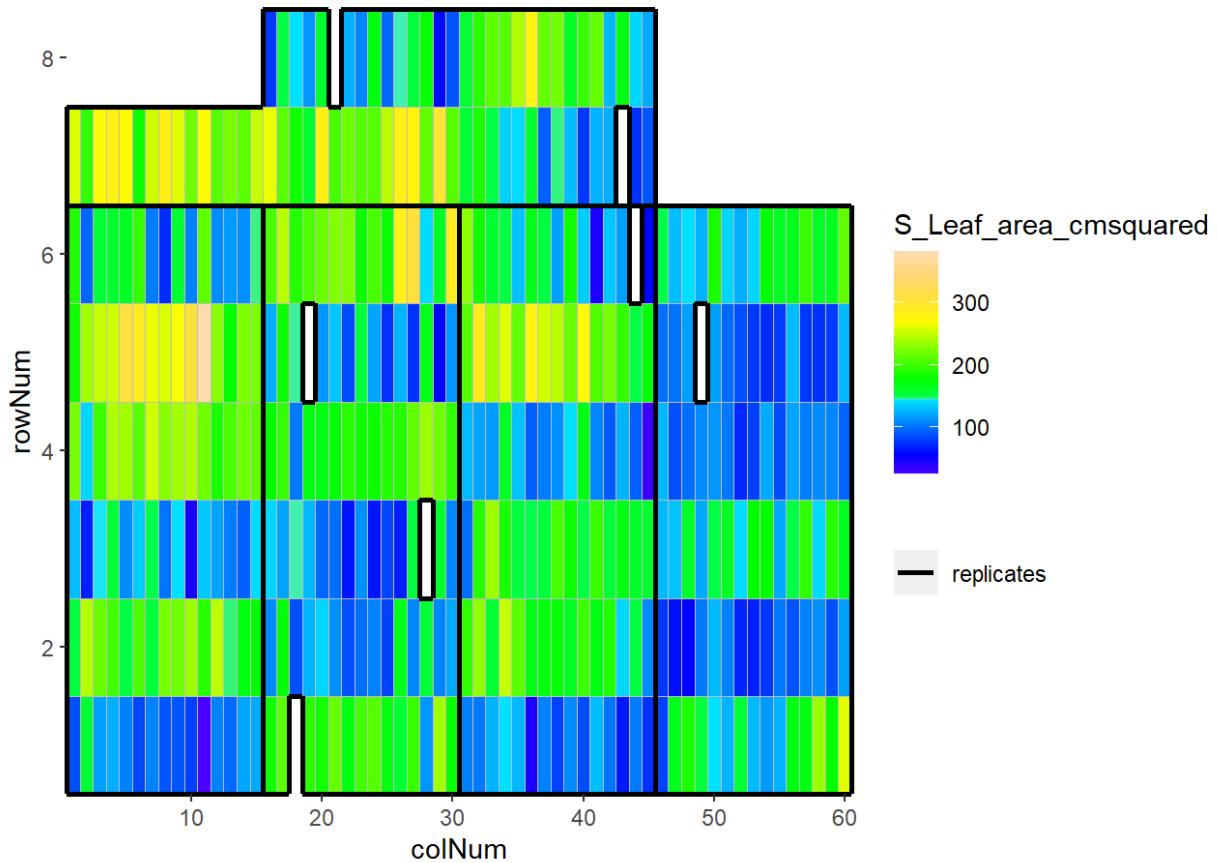
## EPPN2020\_M3P - 2020-03-20



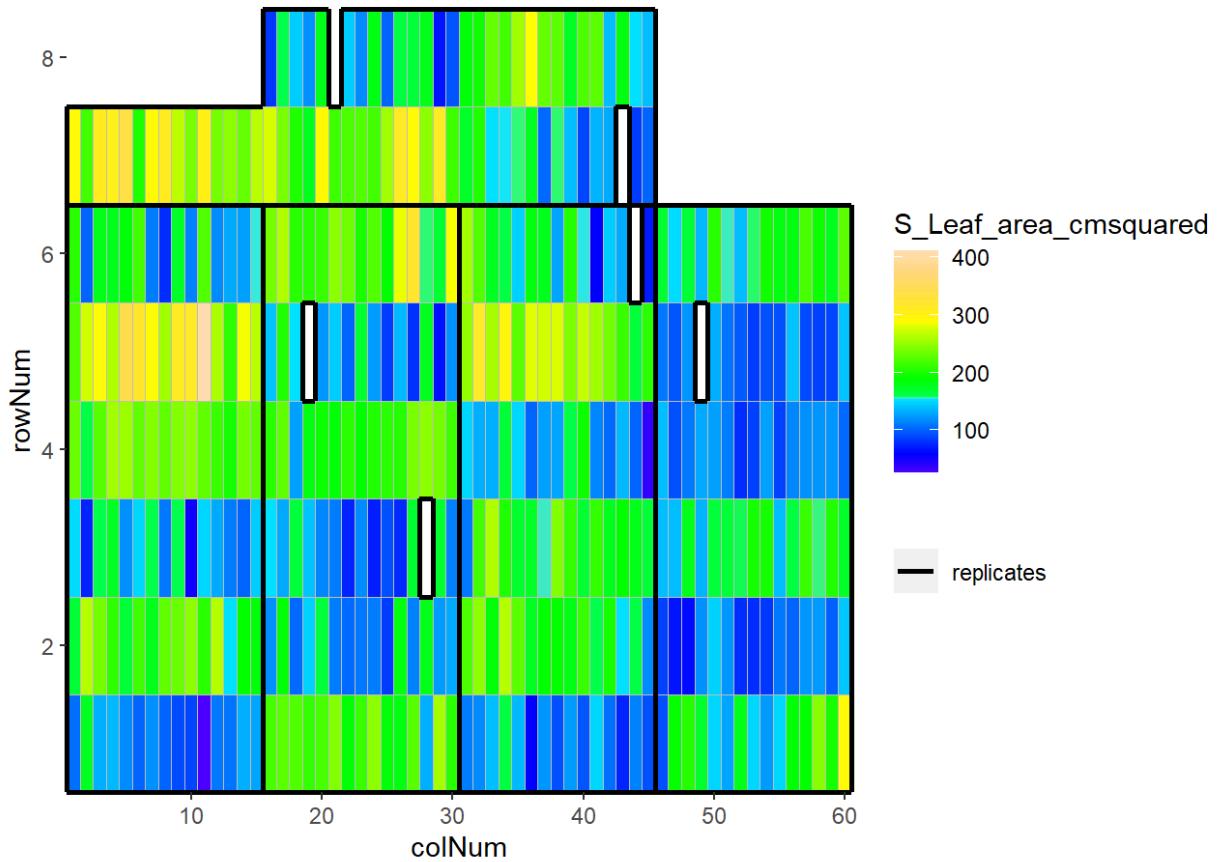
## EPPN2020\_M3P - 2020-03-21



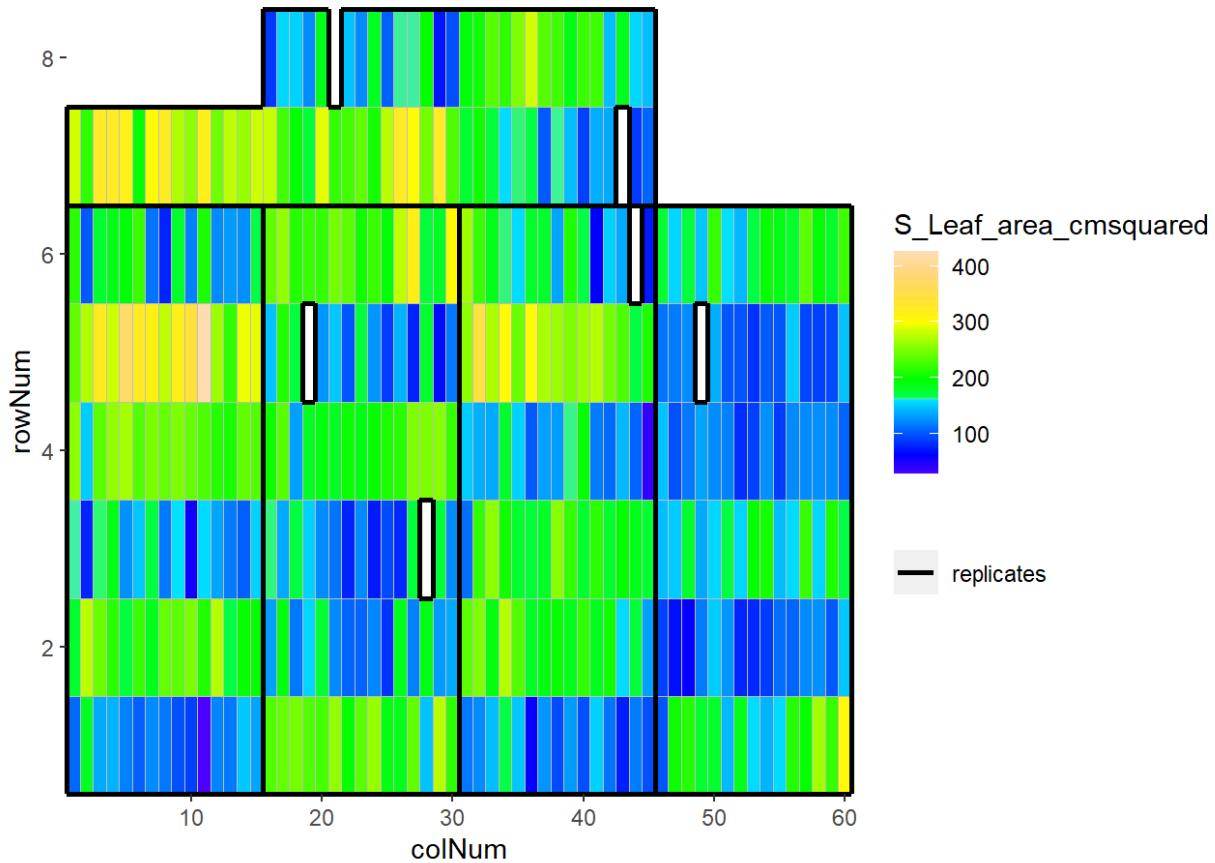
## EPPN2020\_M3P - 2020-03-22



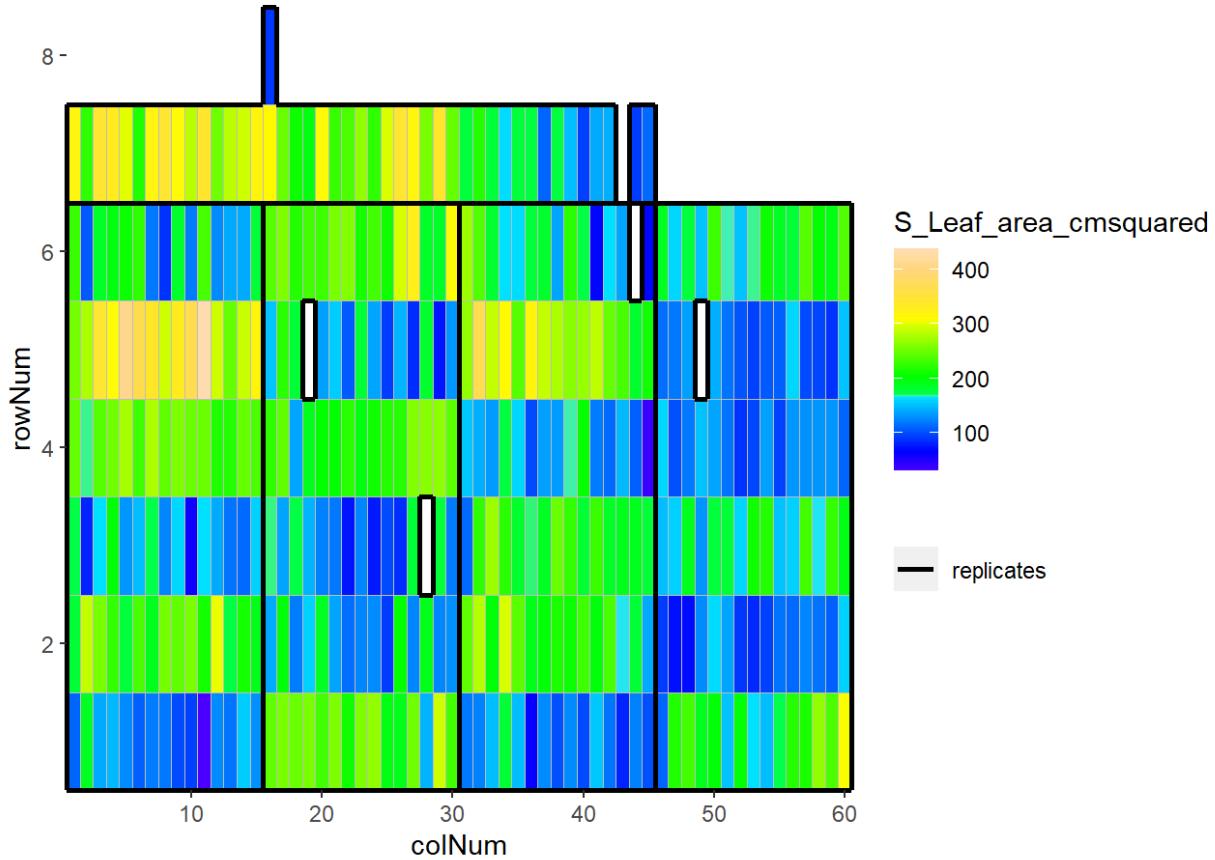
## EPPN2020\_M3P - 2020-03-24



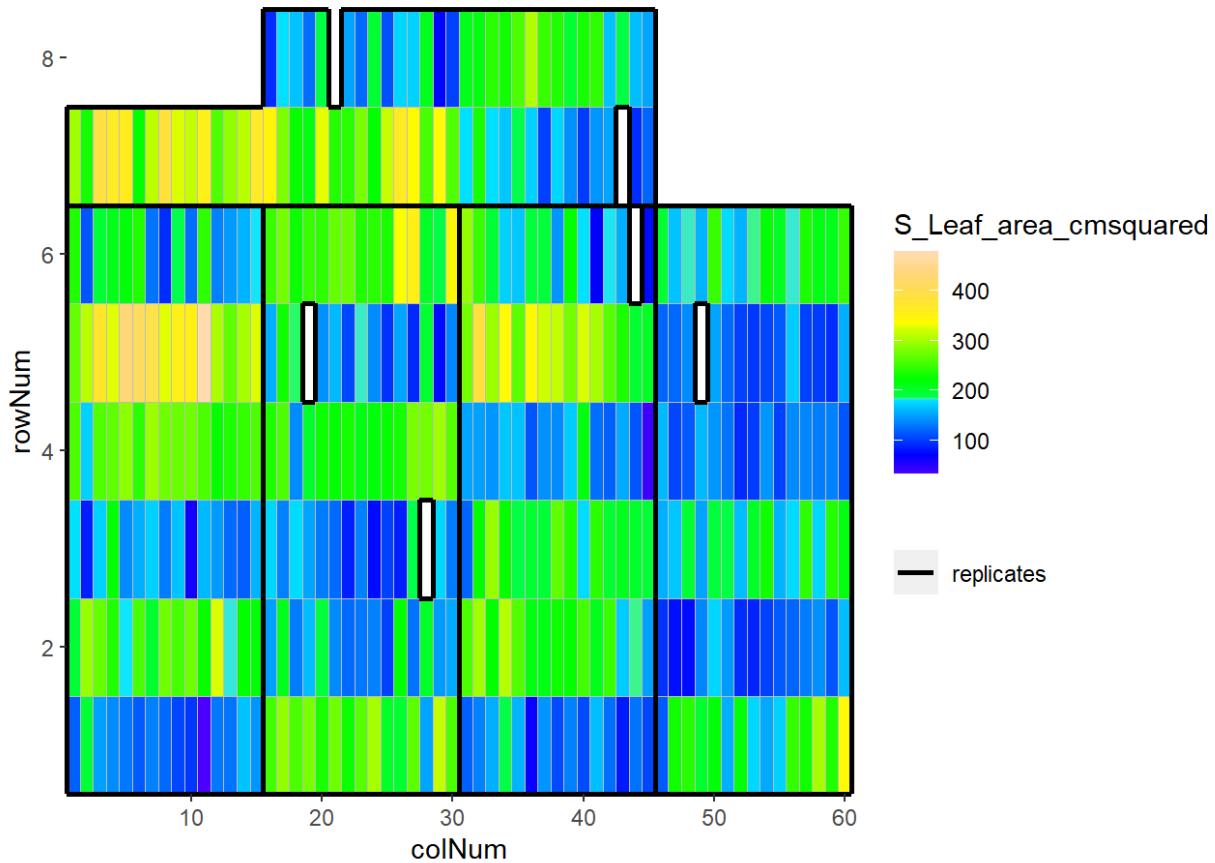
## EPPN2020\_M3P - 2020-03-25



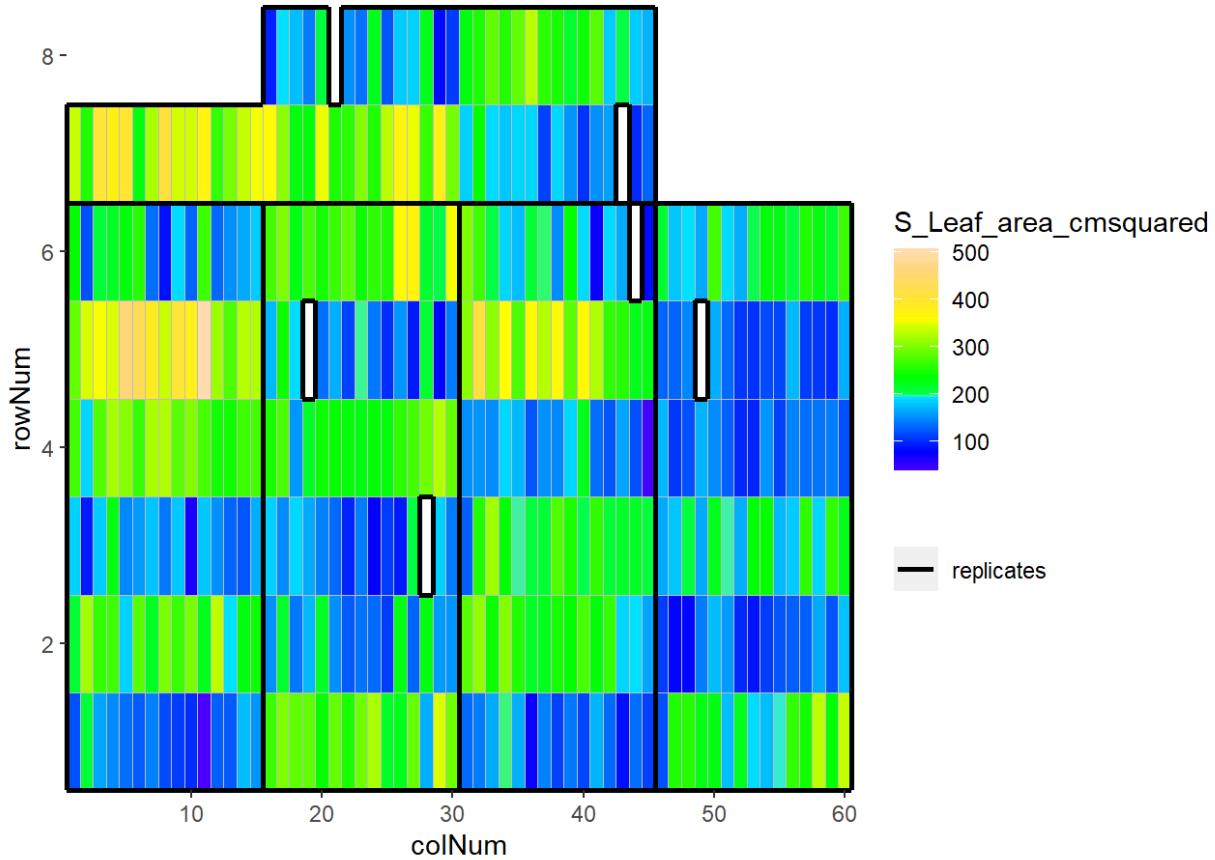
## EPPN2020\_M3P - 2020-03-26



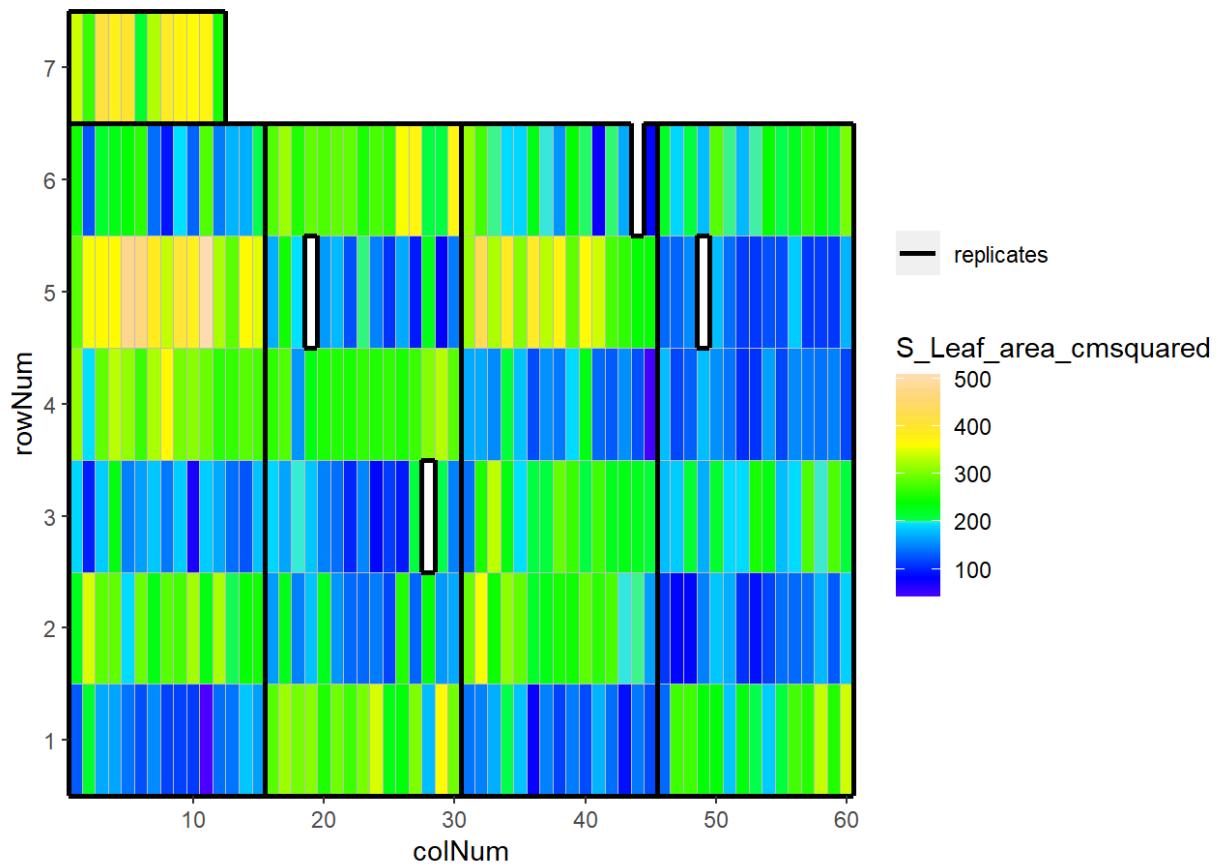
## EPPN2020\_M3P - 2020-03-27



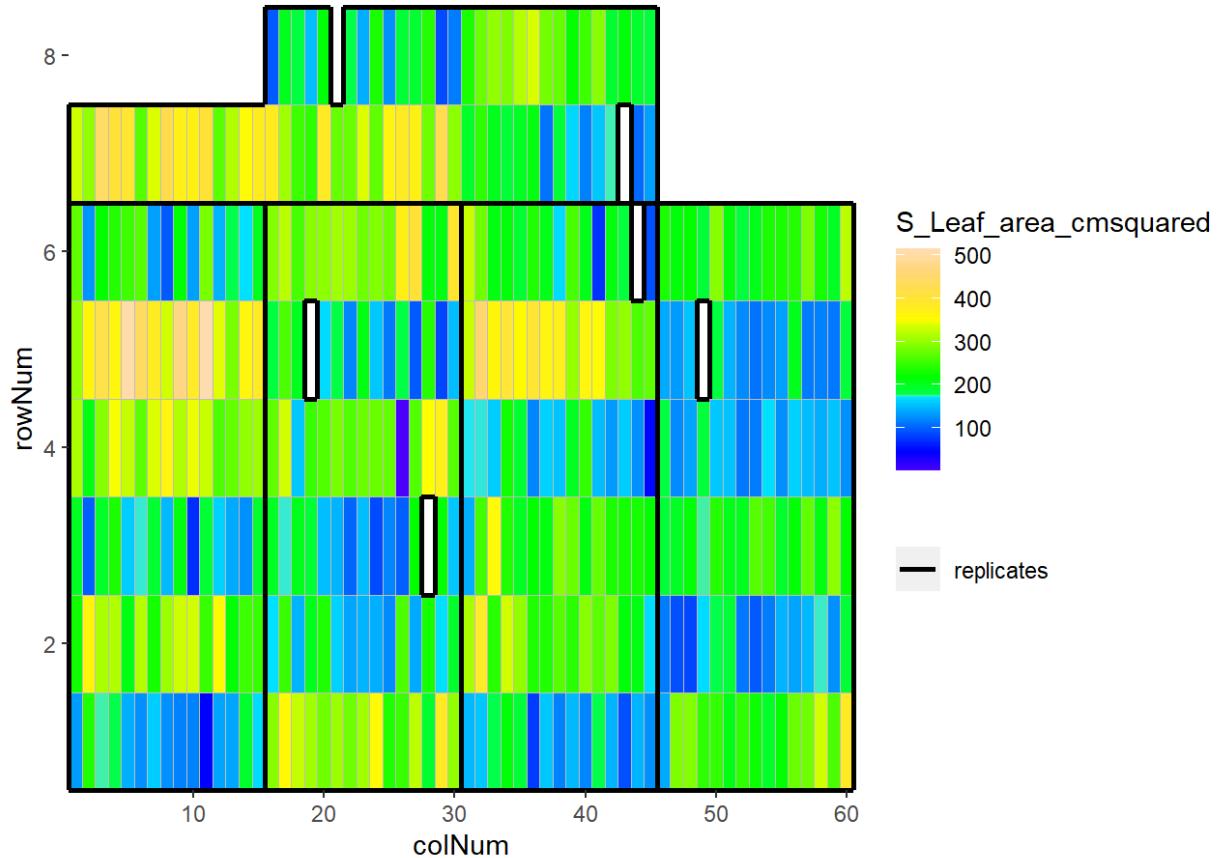
## EPPN2020\_M3P - 2020-03-28



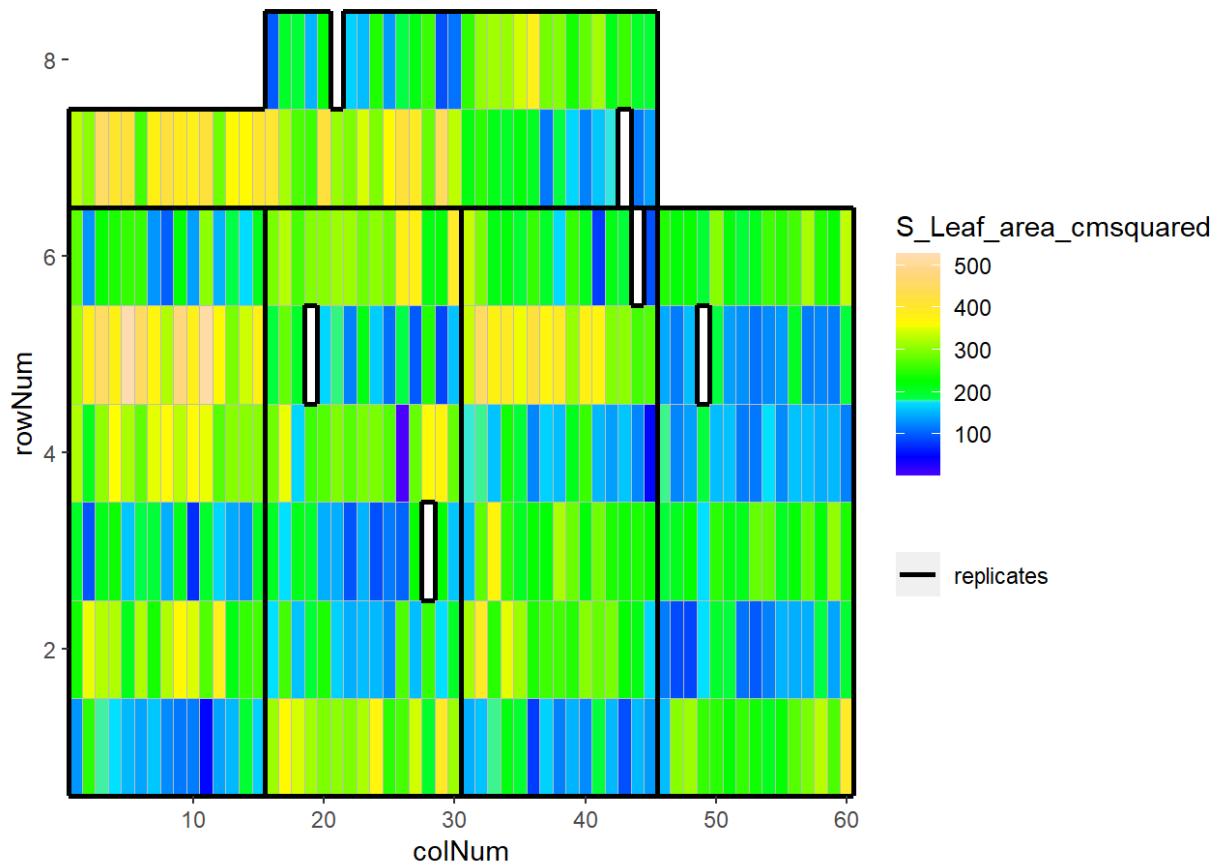
## EPPN2020\_M3P - 2020-03-29



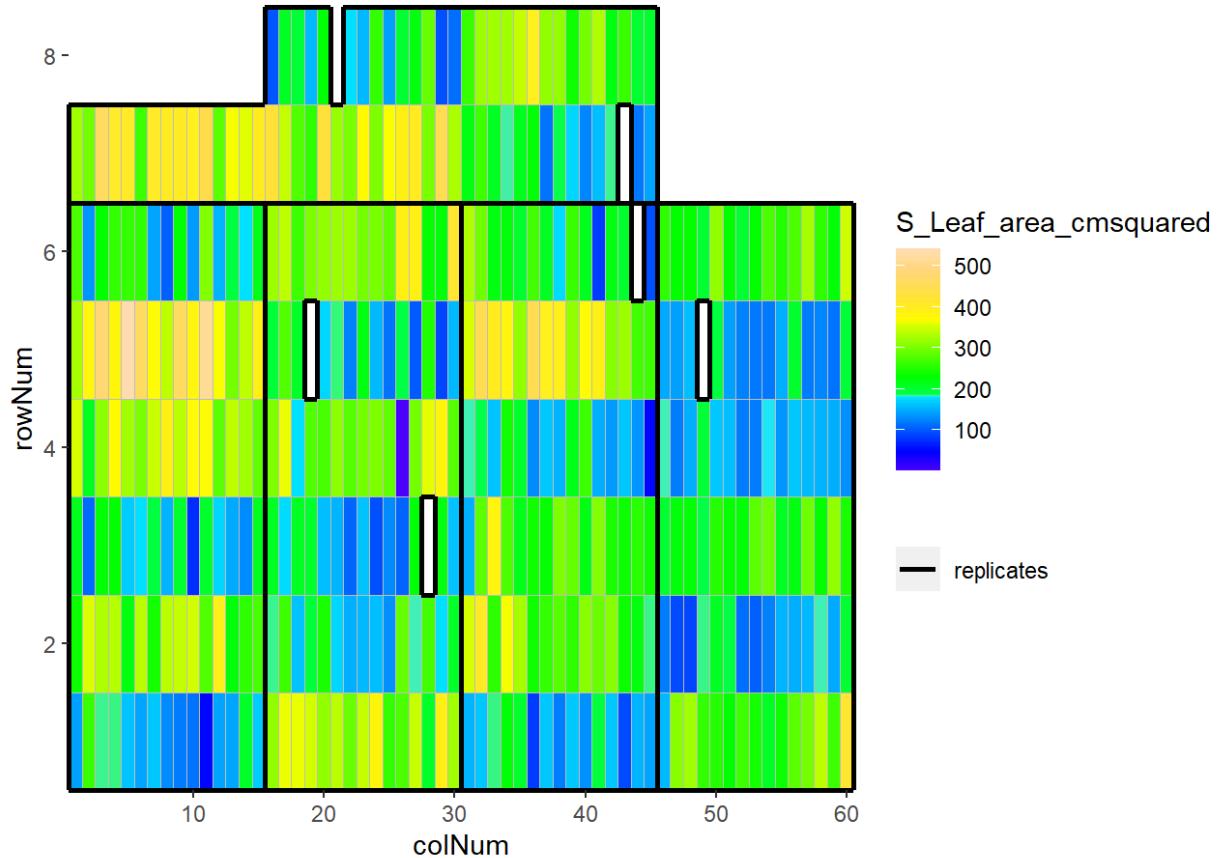
## EPPN2020\_M3P - 2020-03-30



## EPPN2020\_M3P - 2020-03-31



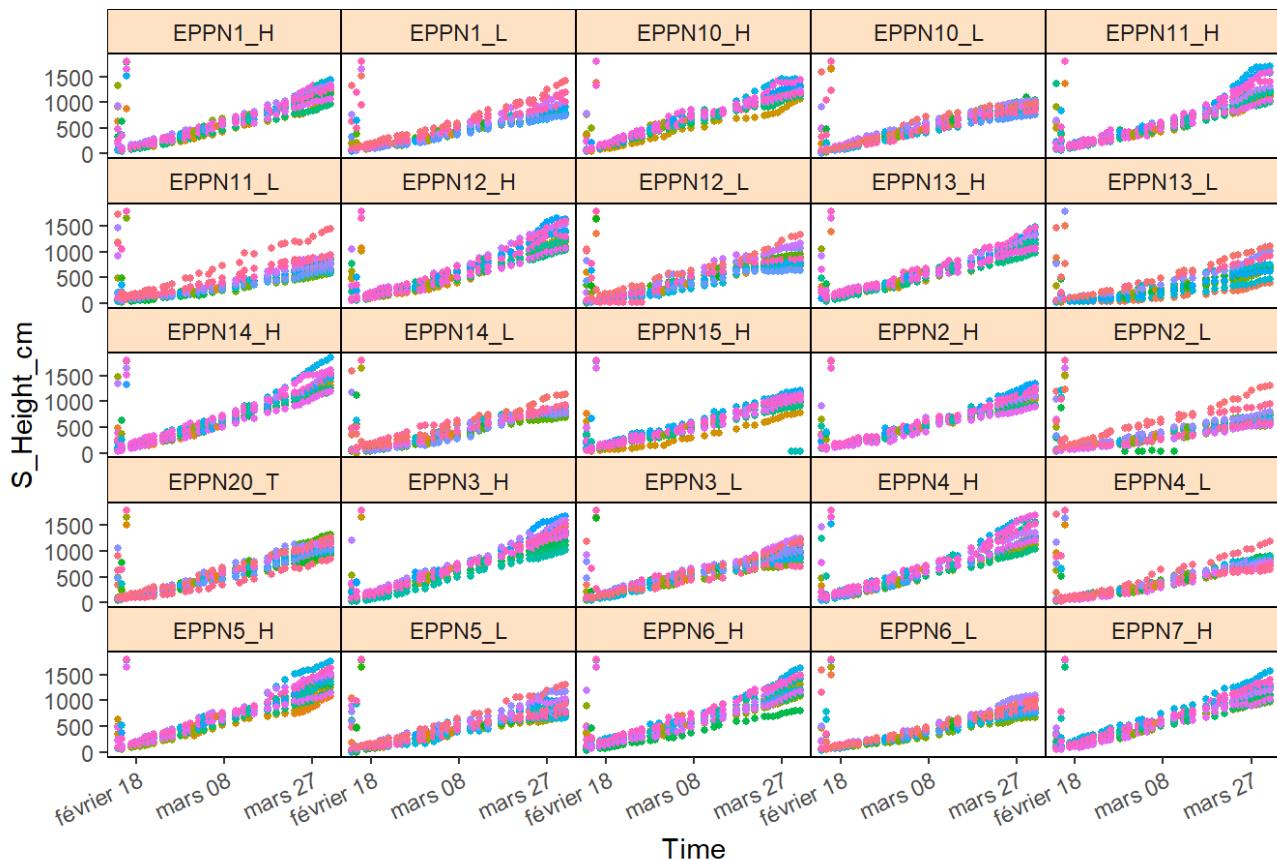
## EPPN2020\_M3P - 2020-04-01



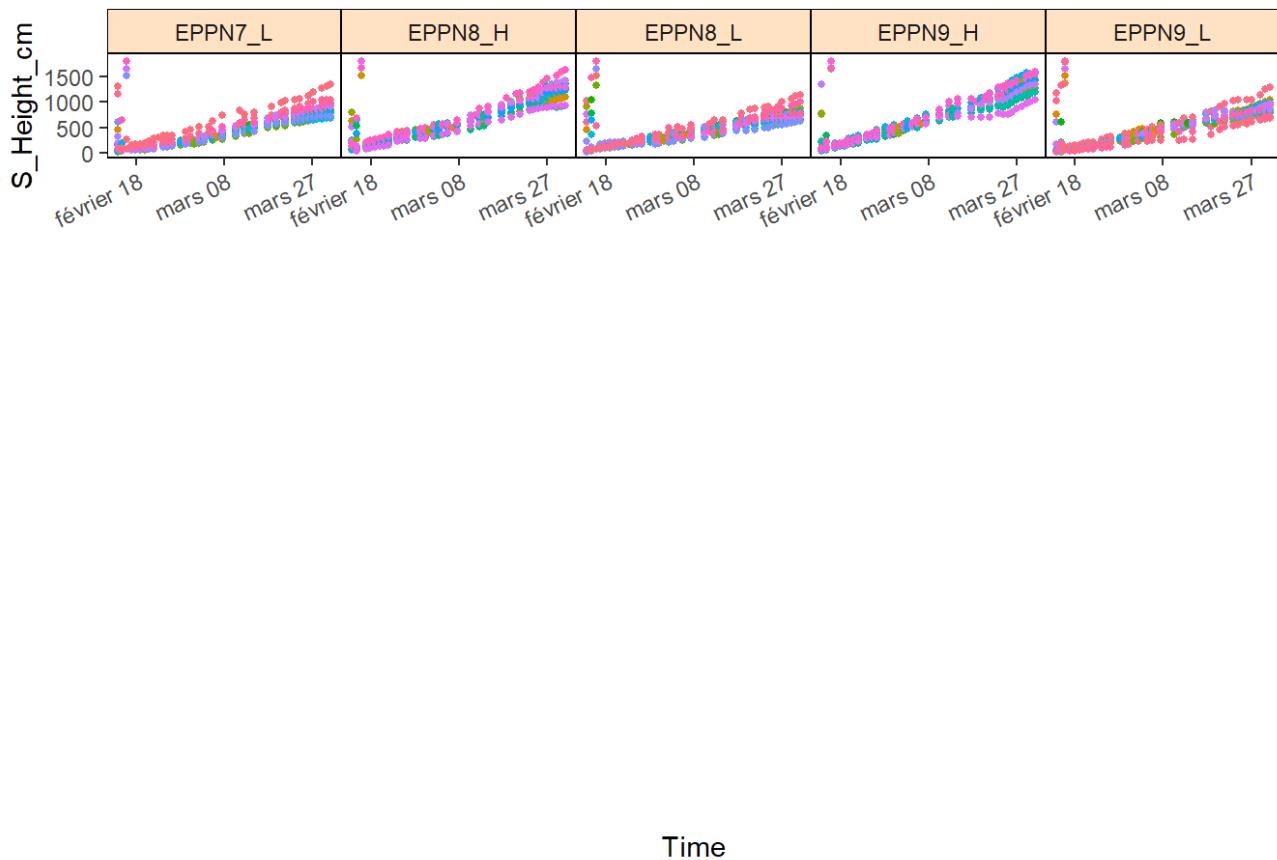
Check time course of raw data per time point

```
for (trait_name in traits) {  
  plot(timePoint_S,  
    traits = trait_name,  
    plotType = "raw")  
}
```

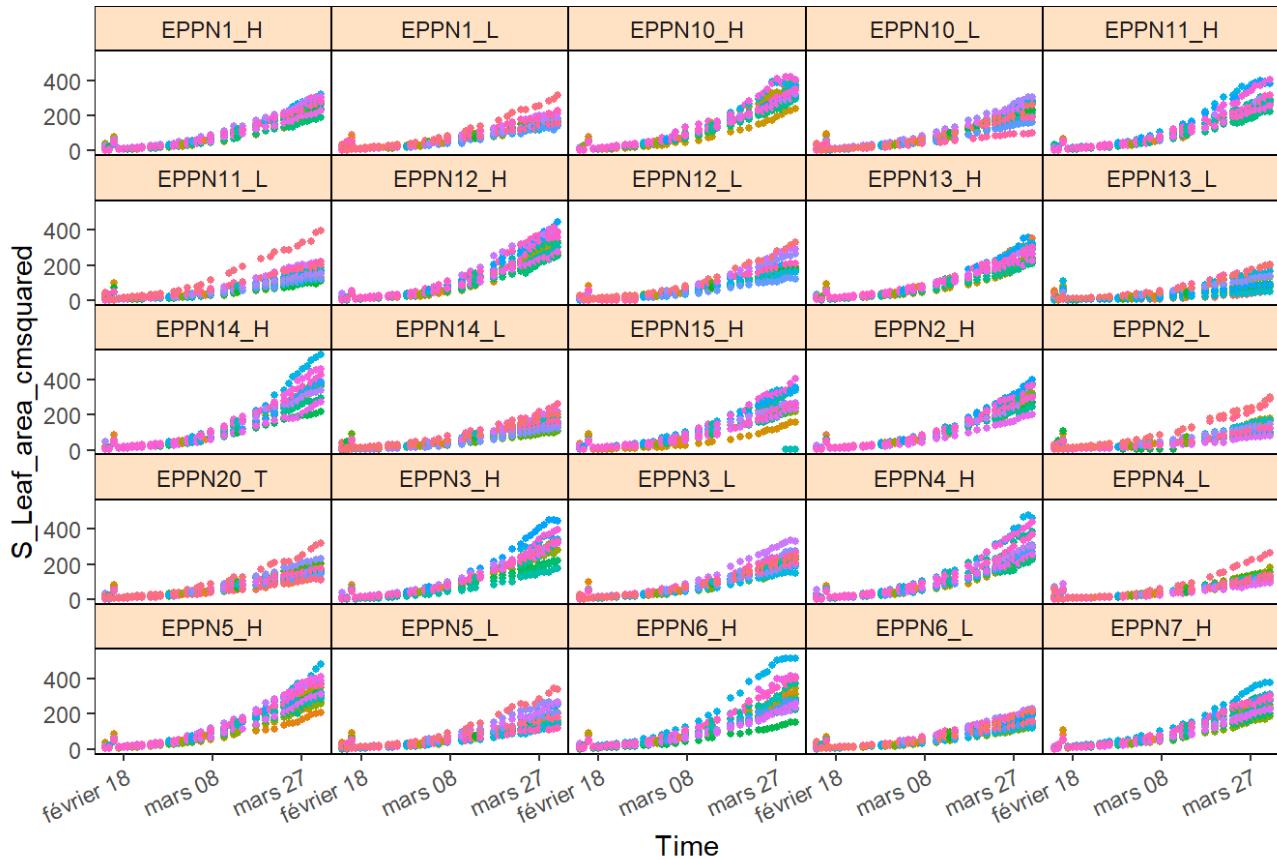
## EPPN2020\_M3P - S\_Height\_cm - raw data



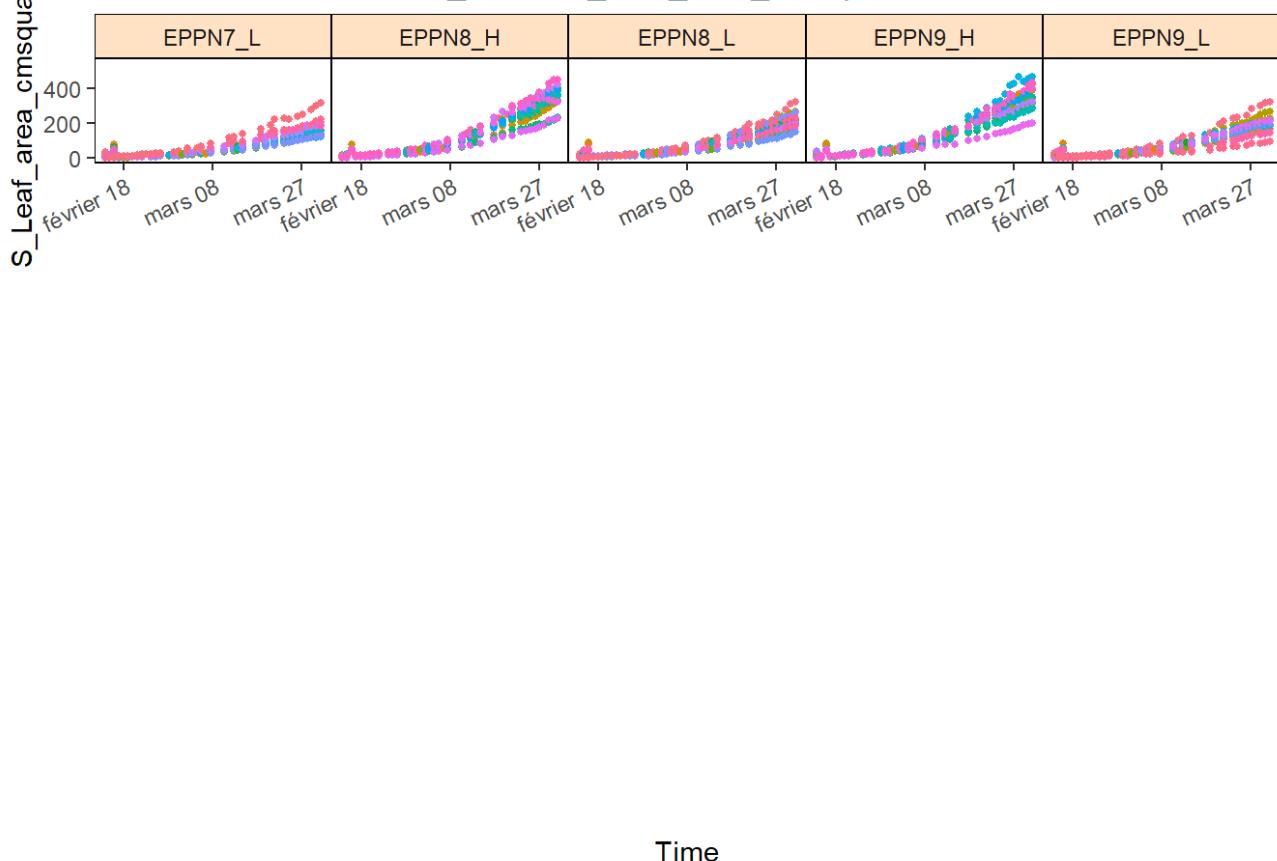
## EPPN2020\_M3P - S\_Height\_cm - raw data



## EPPN2020\_M3P - S\_Leaf\_area\_cmsquared - raw data



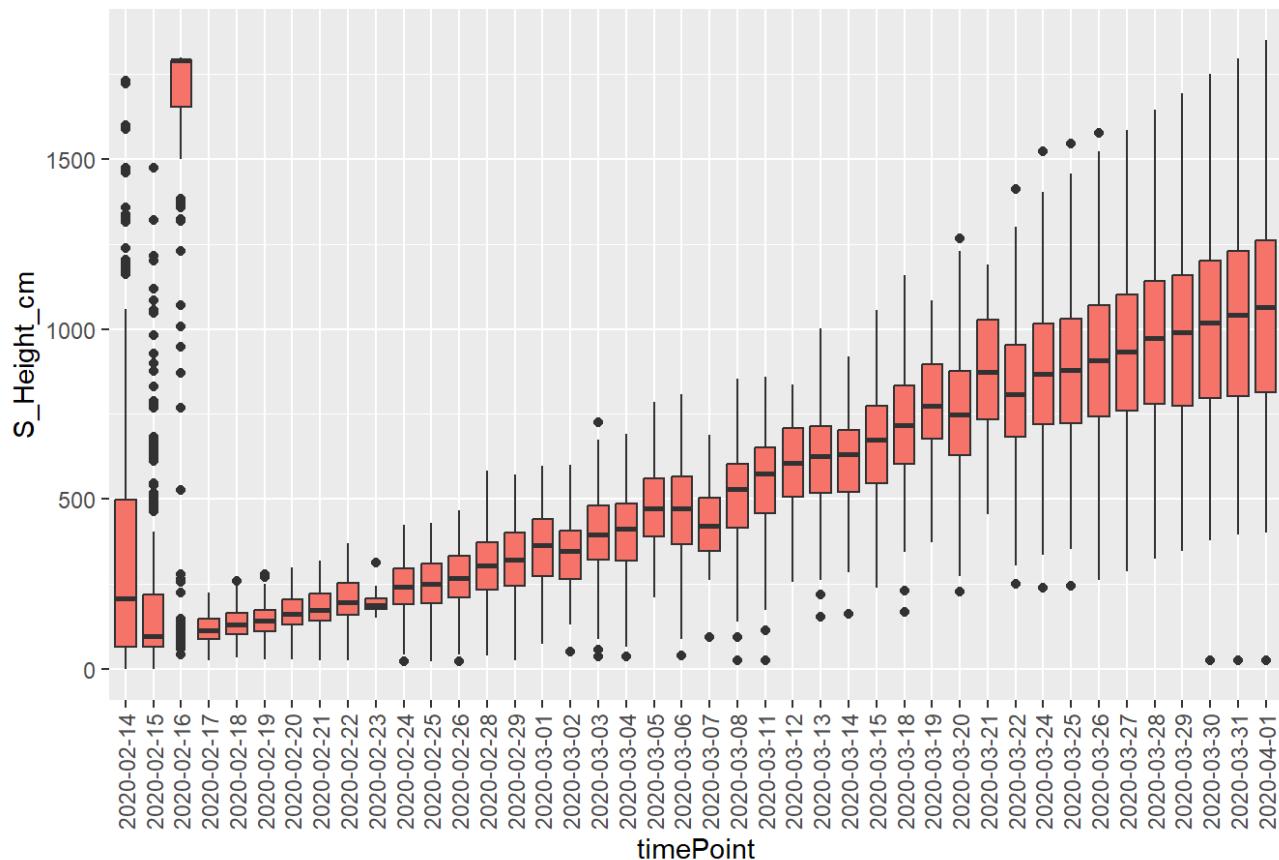
## EPPN2020\_M3P - S\_Leaf\_area\_cmsquared - raw data



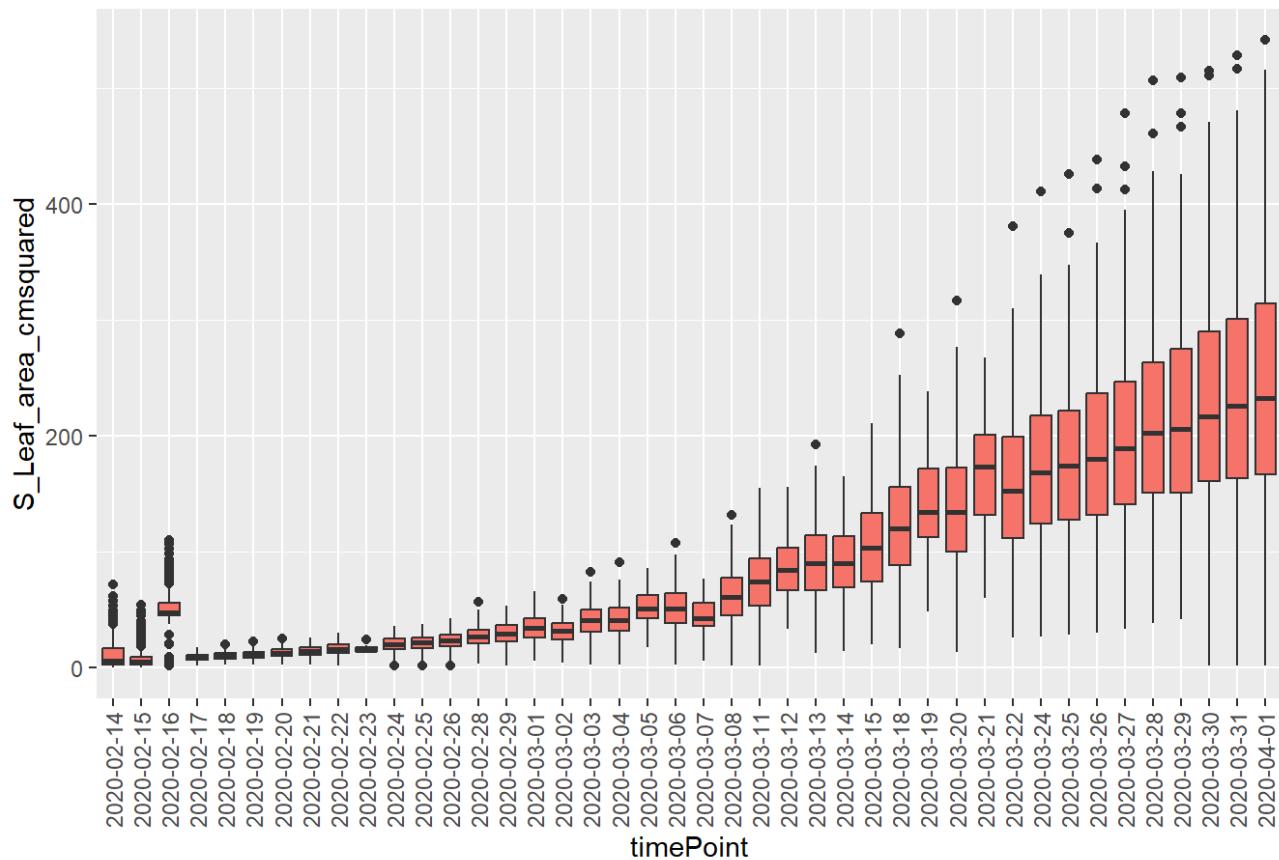
Check the boxplots of raw data per time point

```
for (trait_name in traits) {  
  plot(timePoint_S,  
    plotType = "box",  
    traits = trait_name)  
}
```

## EPPN2020\_M3P - S\_Height\_cm



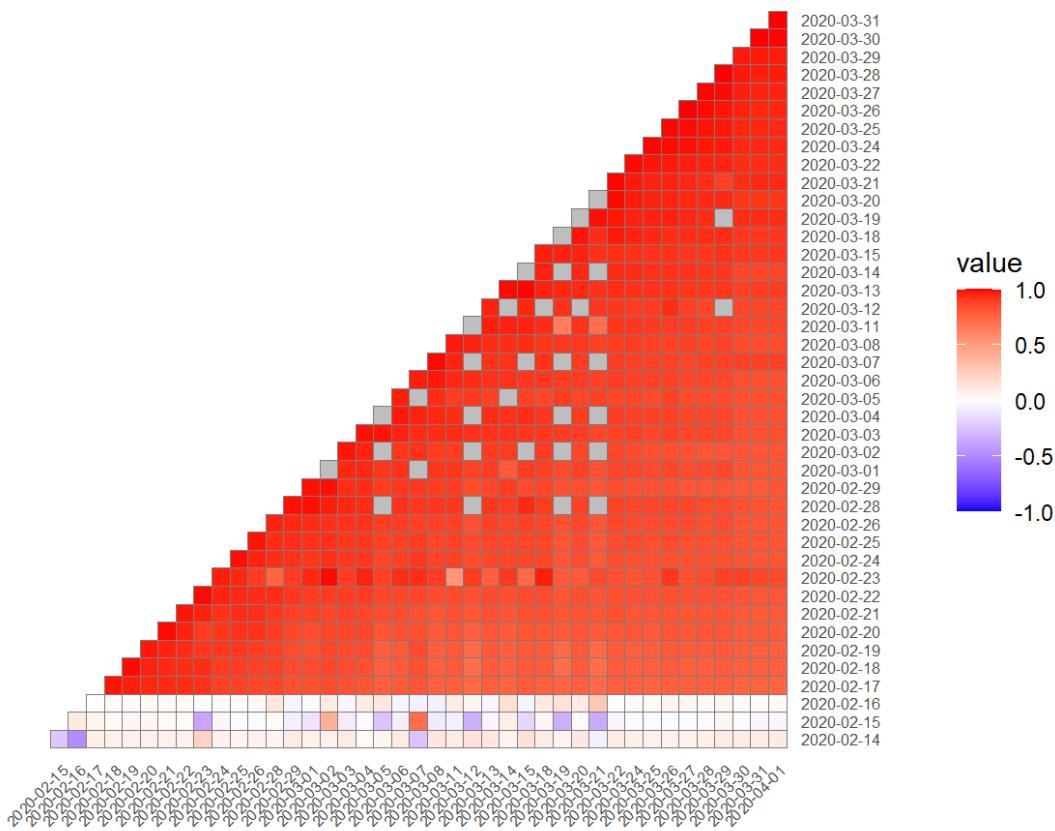
## EPPN2020\_M3P - S\_Leaf\_area\_cmsquared



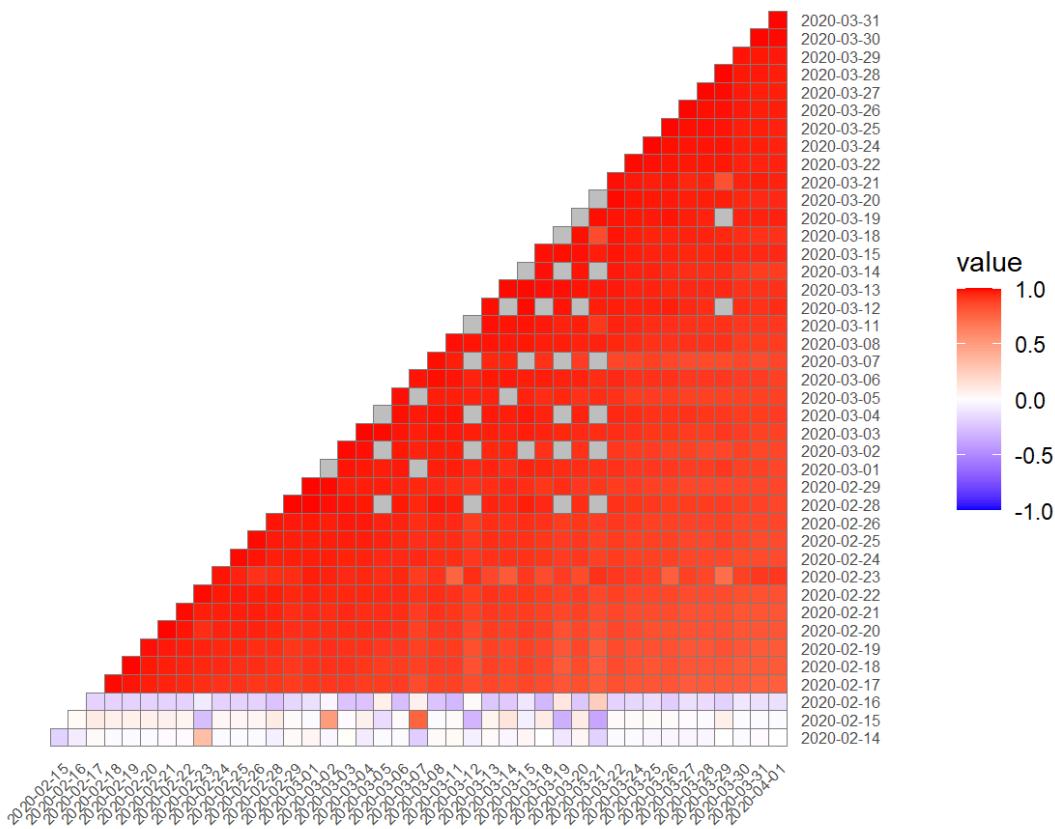
Check the correlation plots of raw data per time point

```
for (trait_name in traits) {  
  plot(timePoint_S,  
    plotType = "cor",  
    traits = trait_name)  
}
```

## EPPN2020\_M3P - Correlations of timepoints for S\_Height\_cm



## ?PN2020\_M3P - Correlations of timepoints for S\_Leaf\_area\_cmsquared



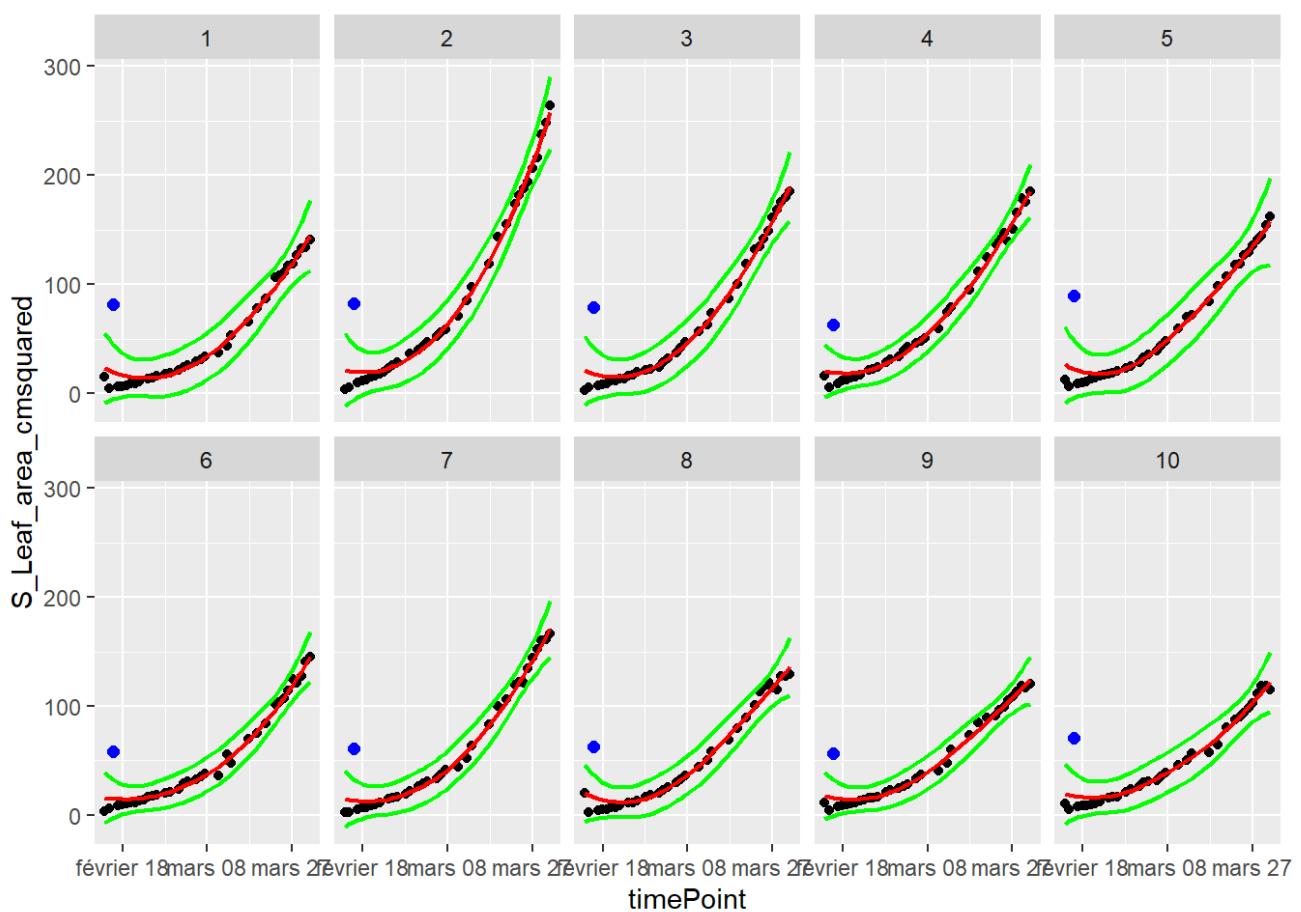
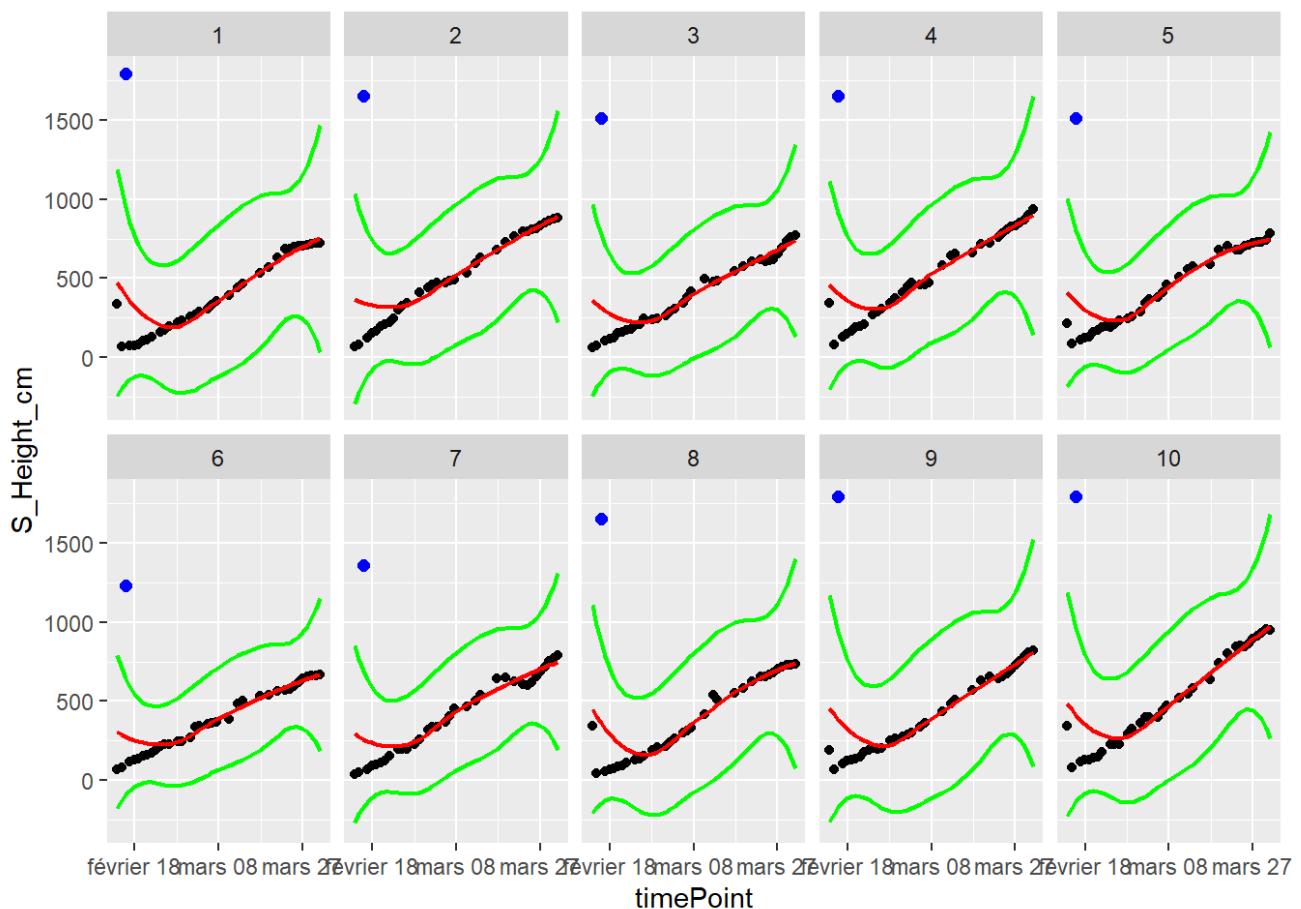
## 1. Detection of outliers for single observations

Using the SingleOut detect and single functions. We select a subset of plants to adjust the settings for the confIntSize and nnLocfit.

```
plantSel<- c(1,2,3,4,5,6,7,8,9,10)
```

```
ci <- 5 # confidence interval  
nn <- 0.8 # nearest neighbor  
ce <- FALSE
```

```
for (trait_name in traits) {  
  variable_name <- paste0("Single_test_", trait_name)  
  
  single_test <- detectSingleOut(  
    TP = timePoint_S,  
    trait = trait_name,  
    plotIds = plantSel,  
    confIntSize = ci,  
    nnLocfit = nn,  
    checkEdges = TRUE # check for outlier values in start and end of experiment  
  )  
  
  assign(variable_name, single_test)  
  
  plot(single_test, outOnly = FALSE)  
}
```



We can then run on all plants of the data set.

```

for (trait_name in traits) {
  single_test_object_name <- paste0("Single_test_", trait_name)
  Single_test <- get(single_test_object_name)
  if (any(Single_test$outlier == 1)) {
    outliers_count <- with(Single_test[Single_test$outlier == 1,], table(timePoint))
    print(trait_name)
    print(outliers_count)

    Single_outliers <- removeSingleOut(timePoint_S, Single_test)
    assign(paste0("Single_outliers_", trait_name), Single_outliers)

    readr::write_tsv(Single_test, sprintf("%s/single_outliers_%s.tsv", datadir, trait_name))
  } else {
    cat("No outlier for", trait_name, "\n")
  }
}

```

```

## [1] "S_Height_cm"
## timePoint
## 2020-02-16
##      10
## [1] "S_Leaf_area_cmsquared"
## timePoint
## 2020-02-16
##      10

```

## Data visualisation after single observations outliers removal

### Heatmap of data

Check the heatmap of the data with outliers detection at all the time points.

```

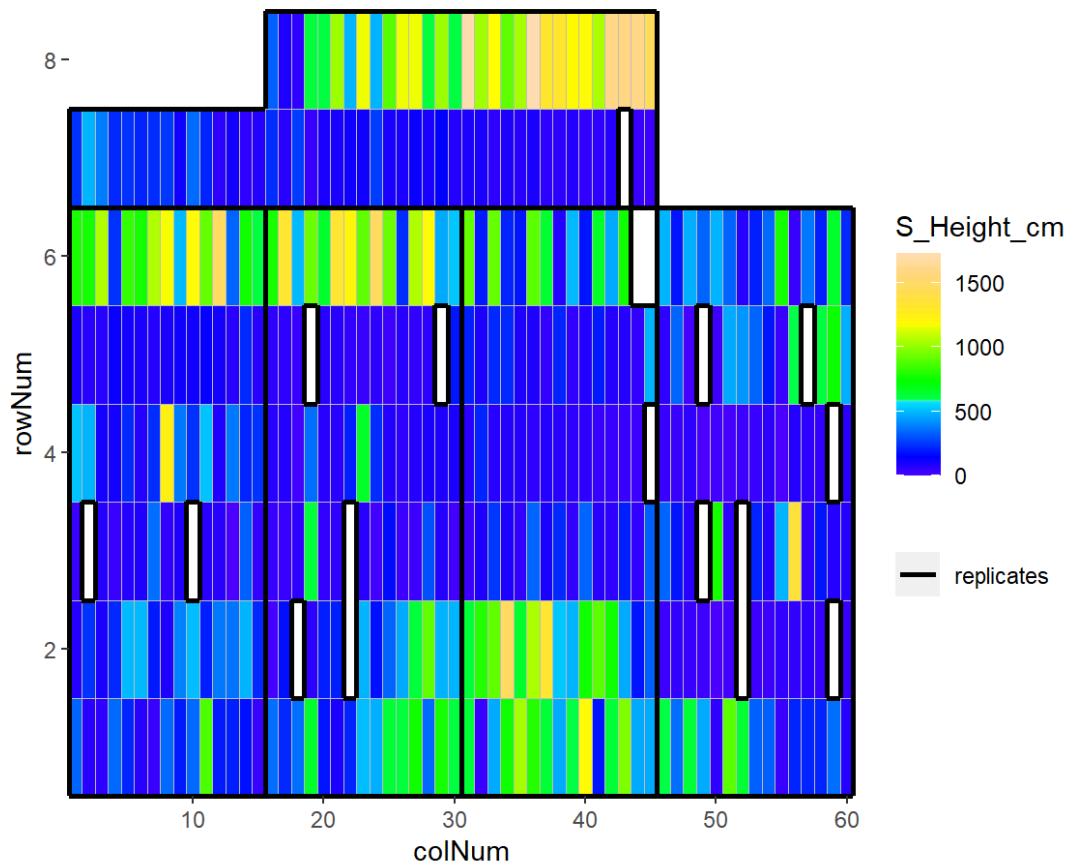
for (trait_name in traits) {
  single_outliers_name <- paste0("Single_outliers_", trait_name)

  if (exists(single_outliers_name)) {
    Single_outliers <- get(single_outliers_name)

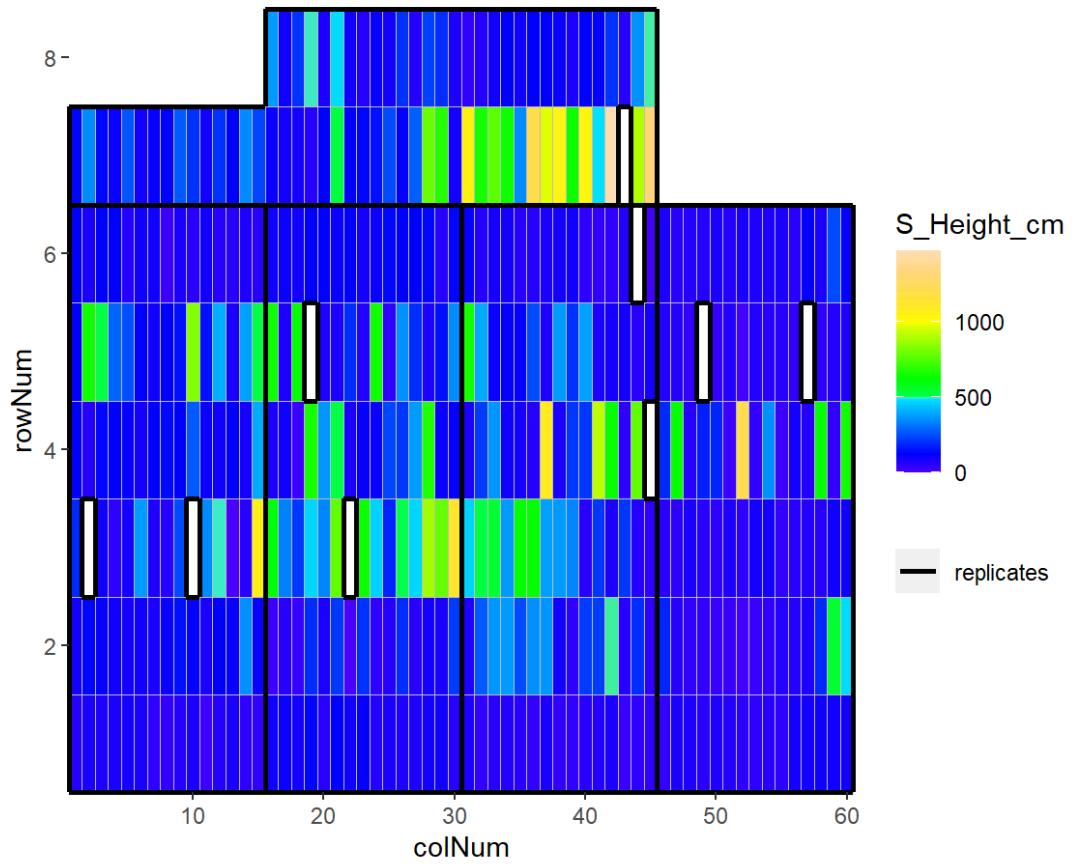
    for (tp in 1:length(num_timepoints$timeNumber)) {
      plot(Single_outliers,
            plotType = "layout",
            timePoints = tp,
            traits = trait_name)
    }
  } else {
    cat("No object Single_outliers found for trait", trait_name, "\n")
  }
}

```

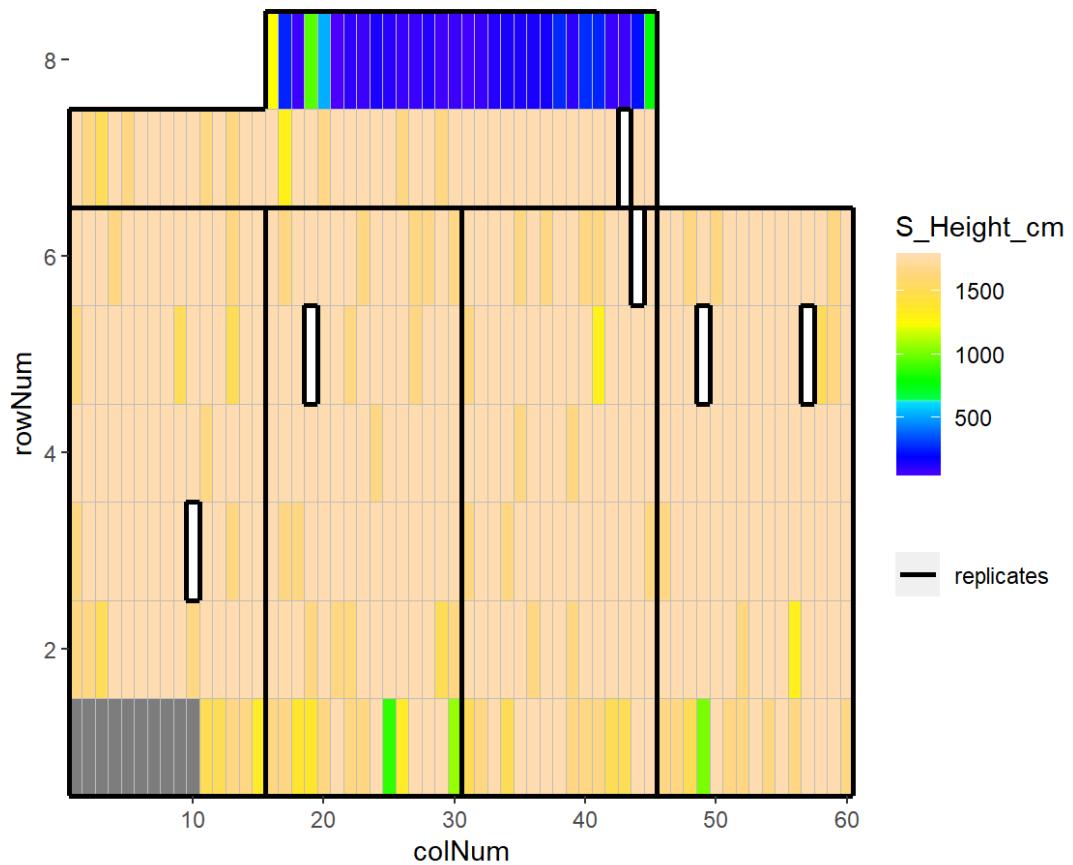
## EPPN2020\_M3P - 2020-02-14



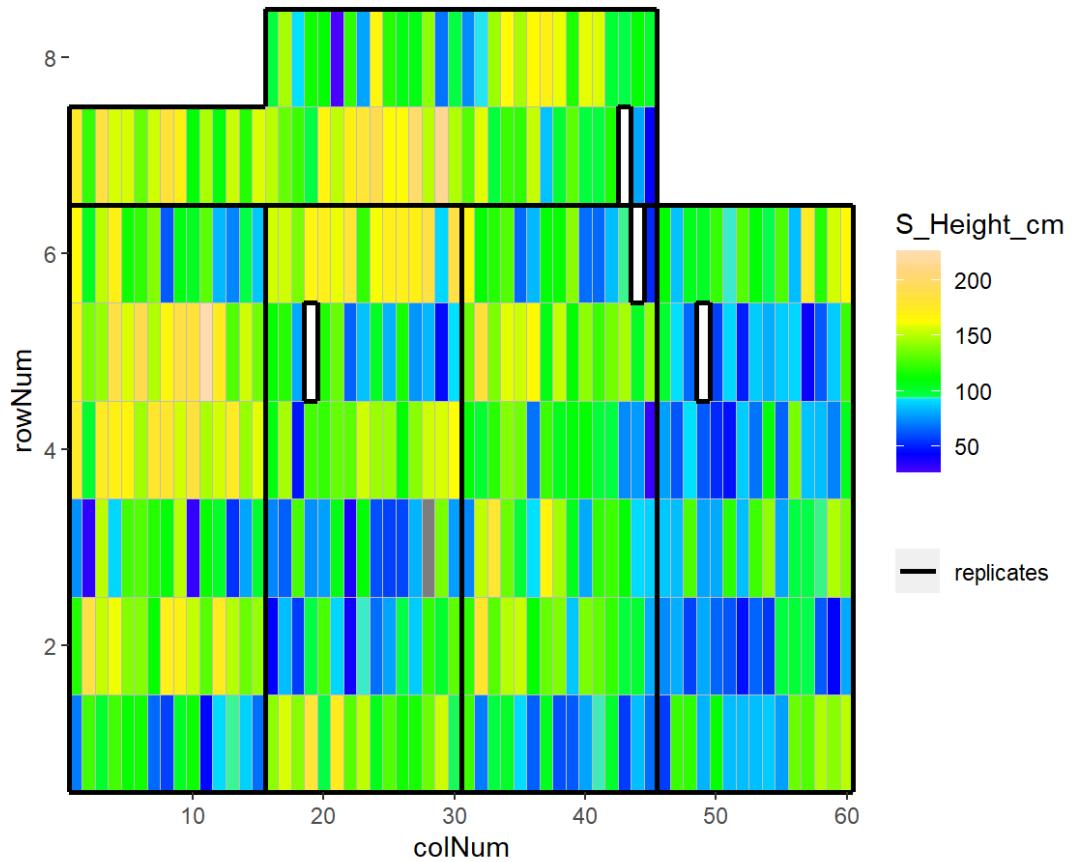
## EPPN2020\_M3P - 2020-02-15



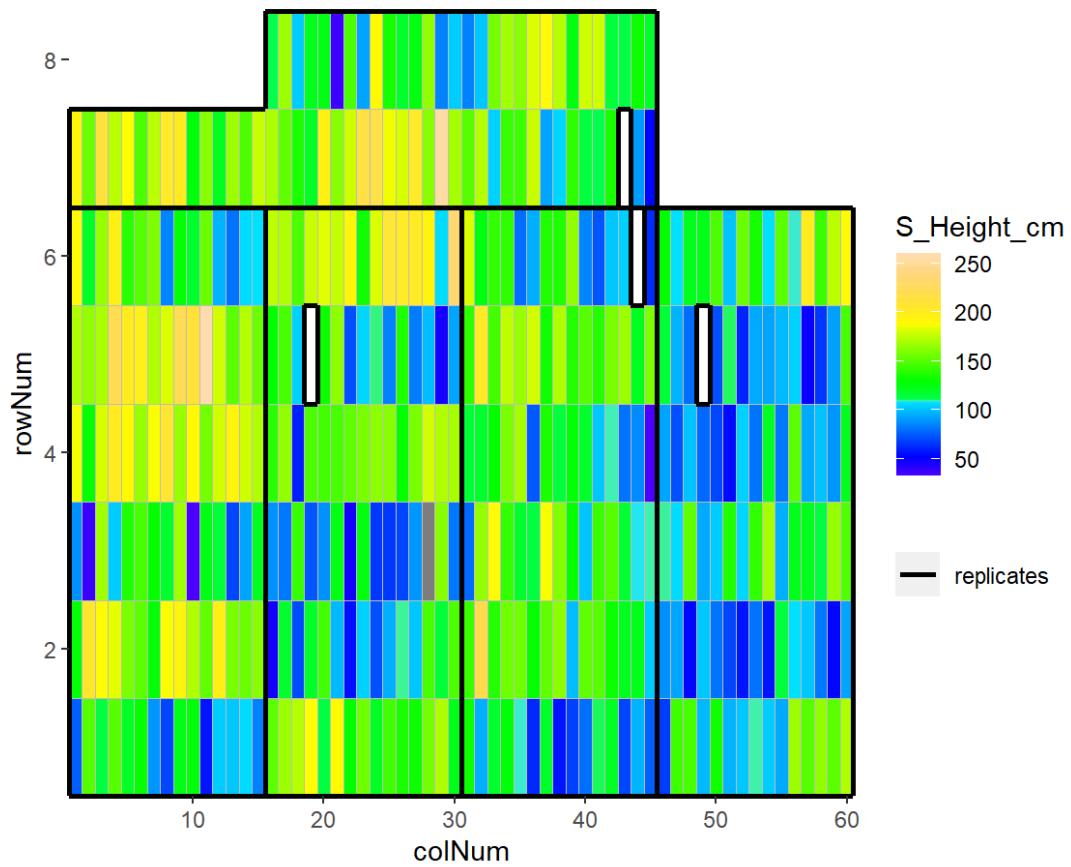
## EPPN2020\_M3P - 2020-02-16



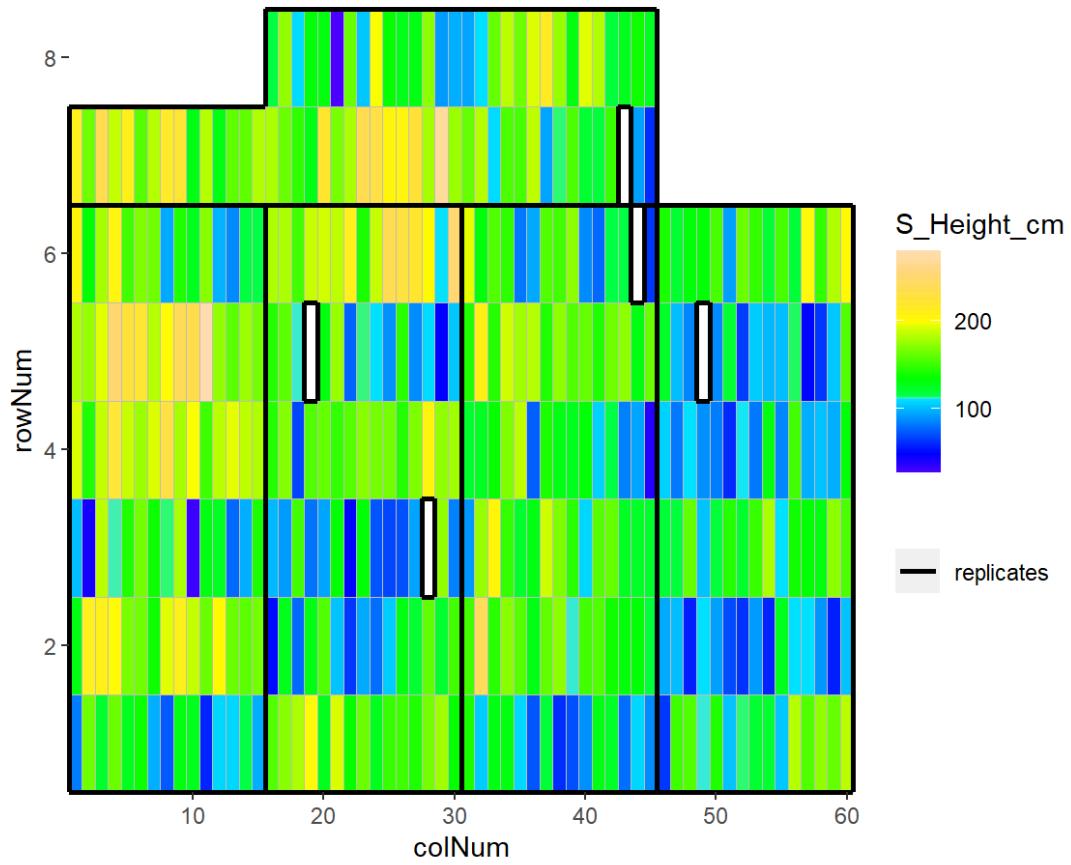
## EPPN2020\_M3P - 2020-02-17



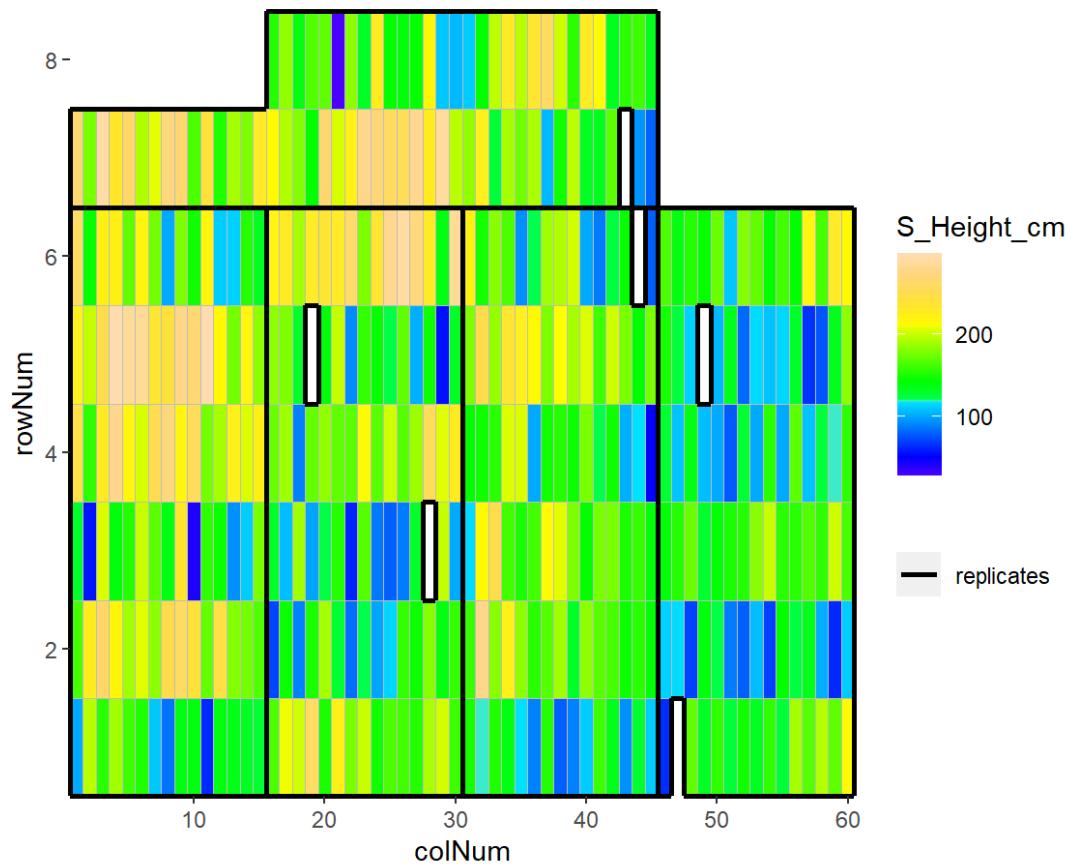
## EPPN2020\_M3P - 2020-02-18



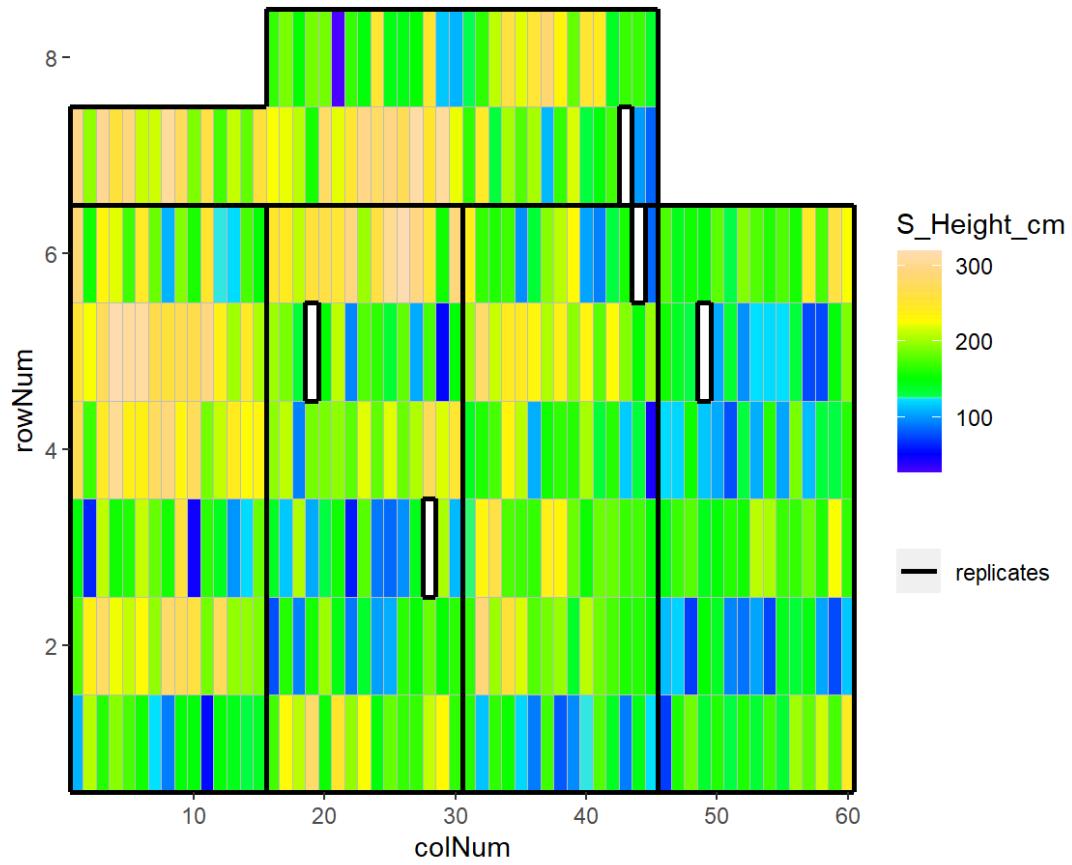
## EPPN2020\_M3P - 2020-02-19



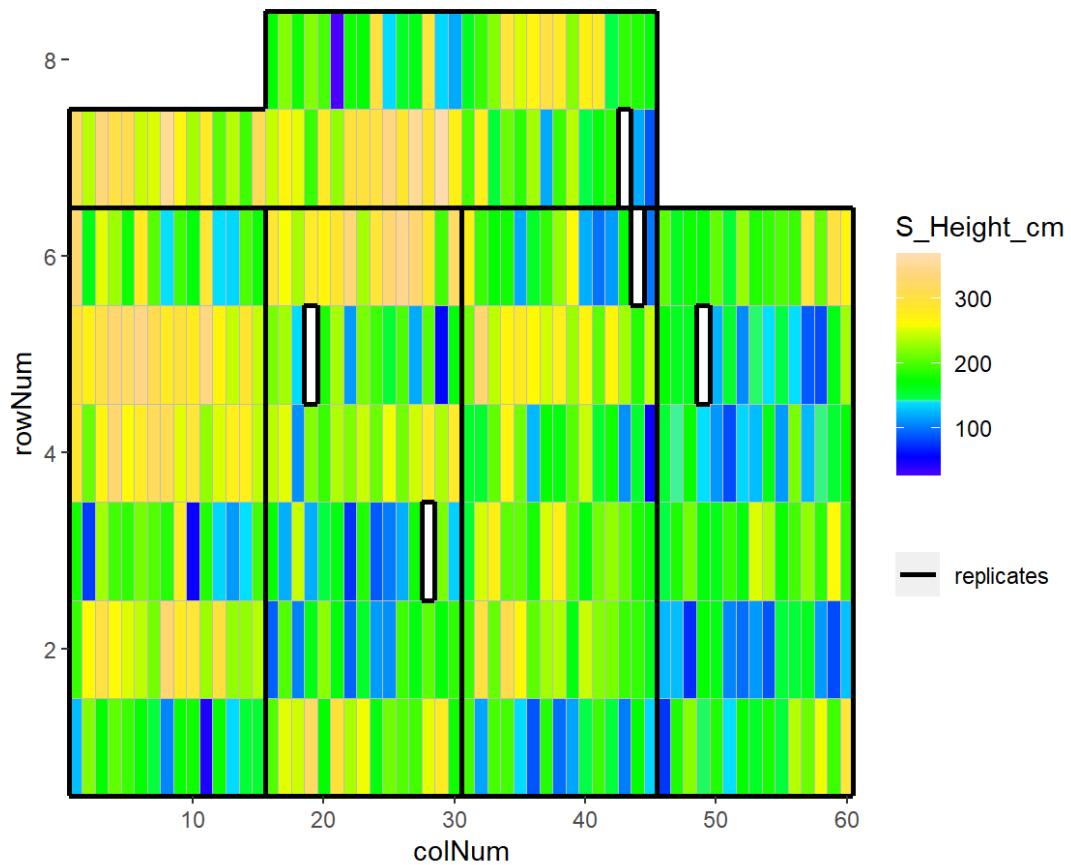
## EPPN2020\_M3P - 2020-02-20



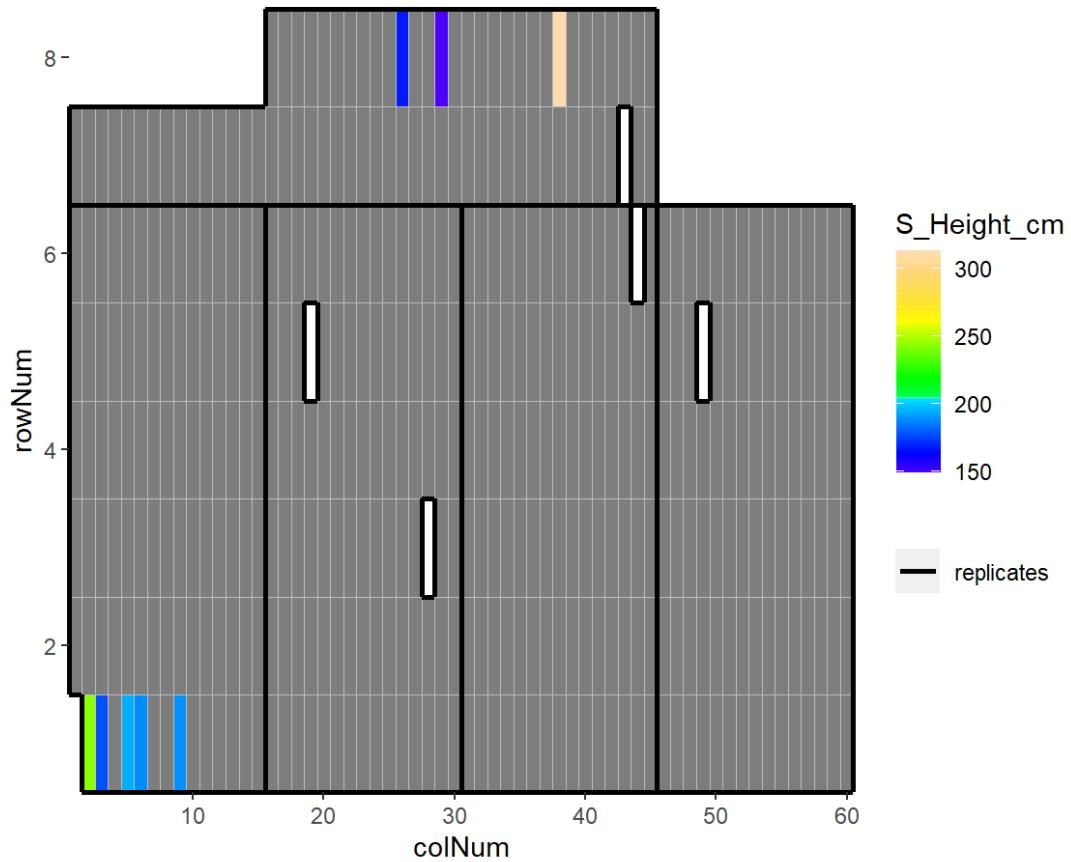
## EPPN2020\_M3P - 2020-02-21



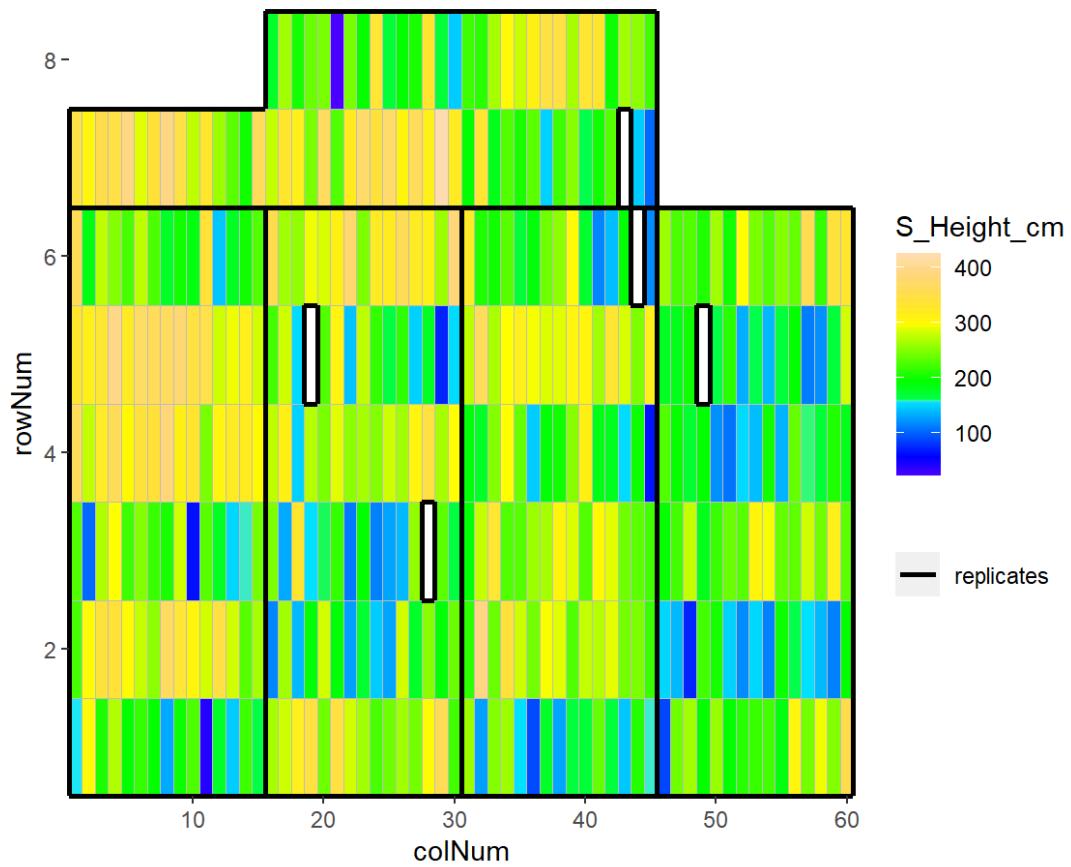
## EPPN2020\_M3P - 2020-02-22



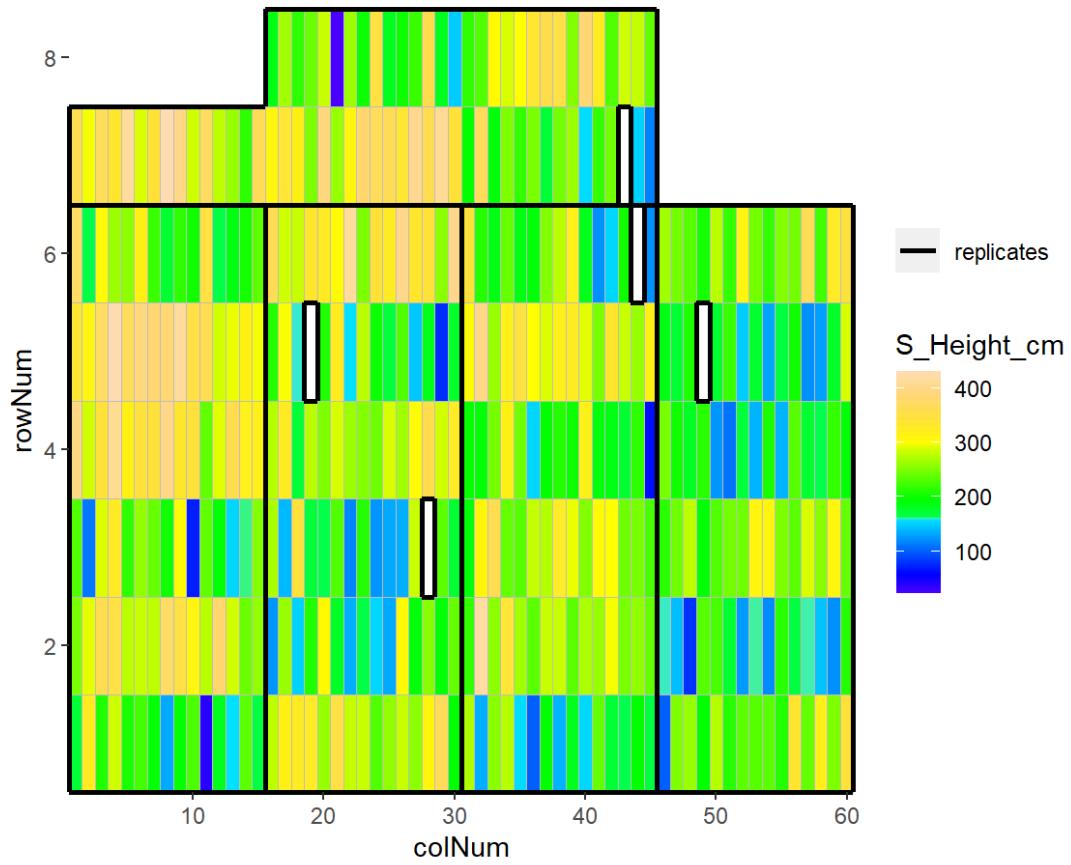
## EPPN2020\_M3P - 2020-02-23



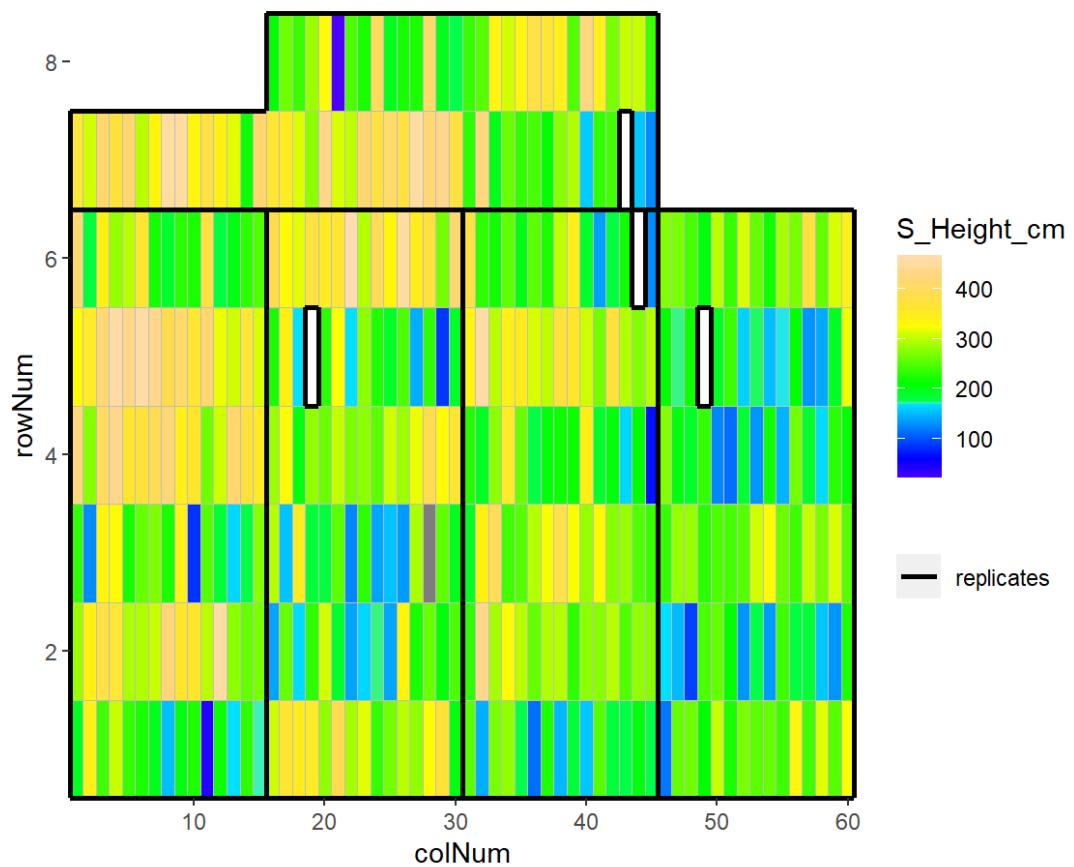
## EPPN2020\_M3P - 2020-02-24



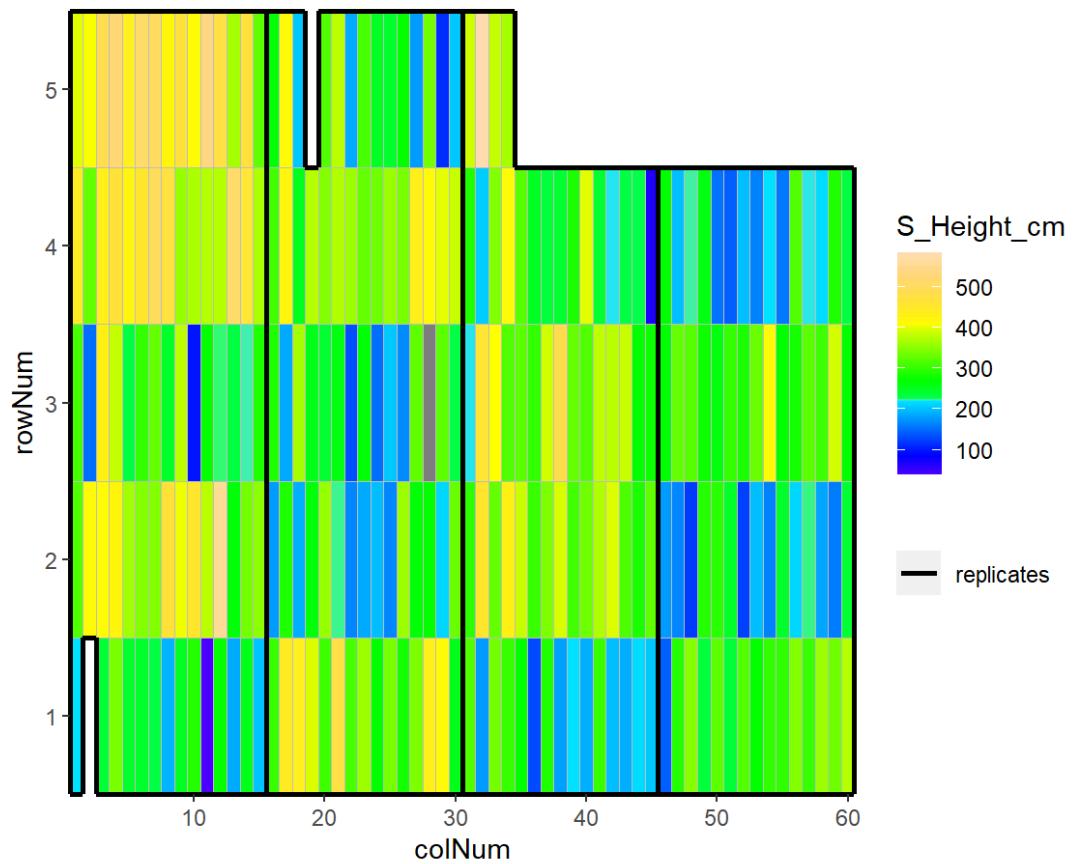
## EPPN2020\_M3P - 2020-02-25



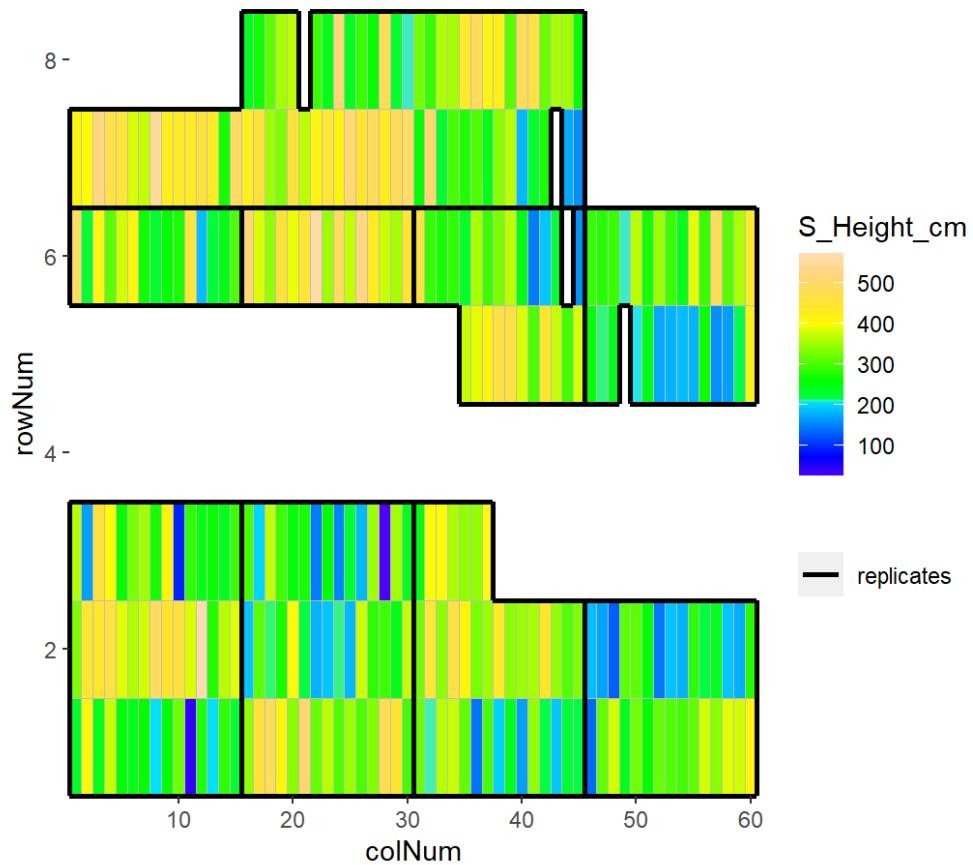
## EPPN2020\_M3P - 2020-02-26



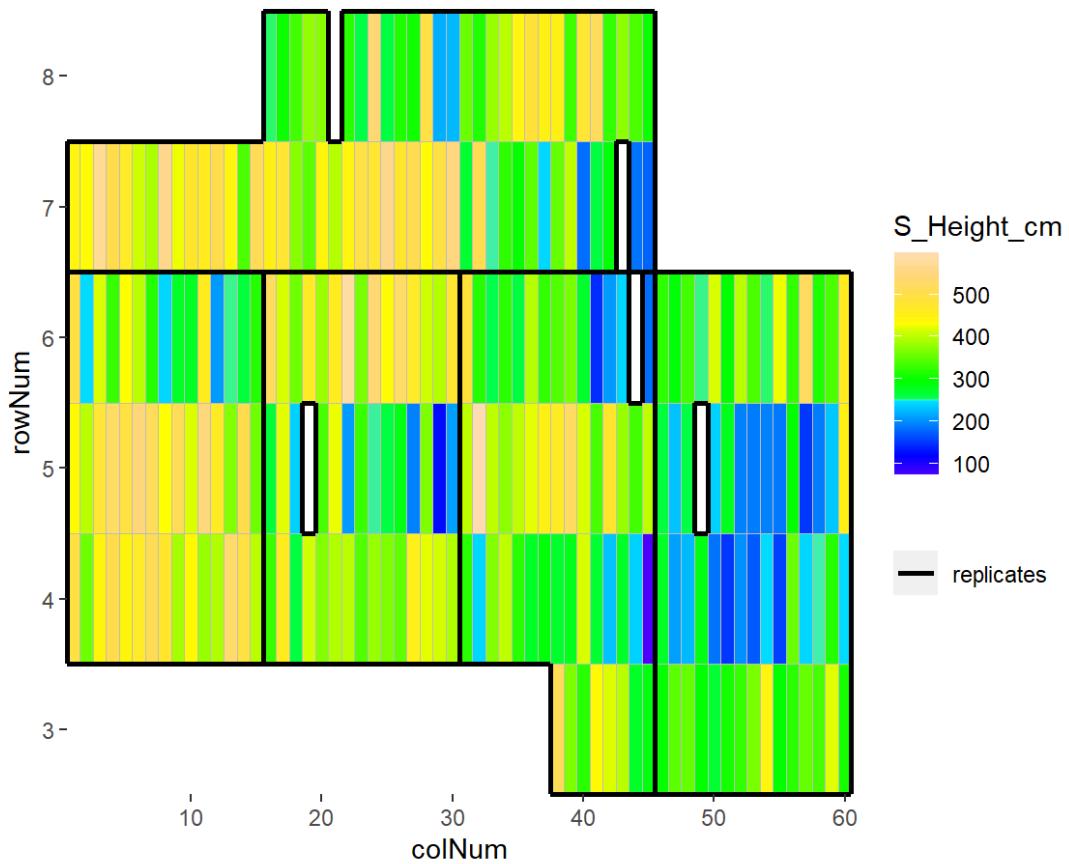
## EPPN2020\_M3P - 2020-02-28



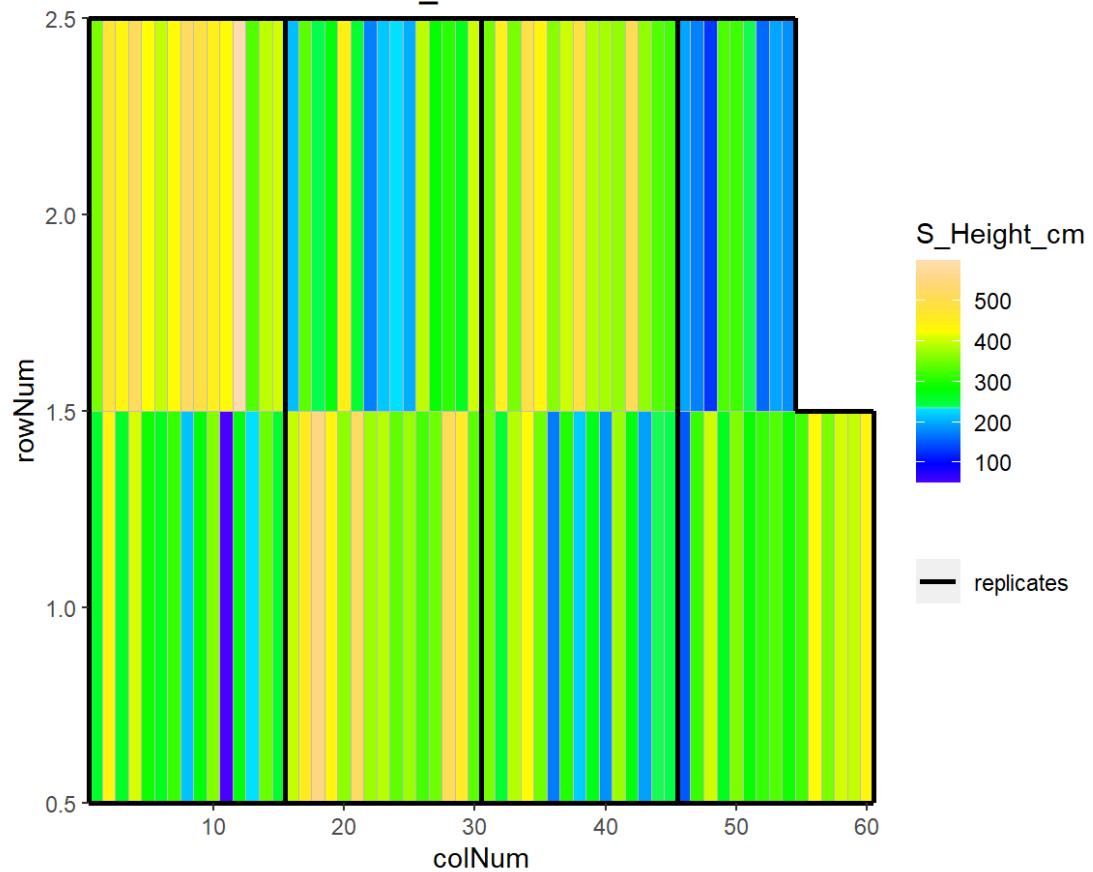
## EPPN2020\_M3P - 2020-02-29



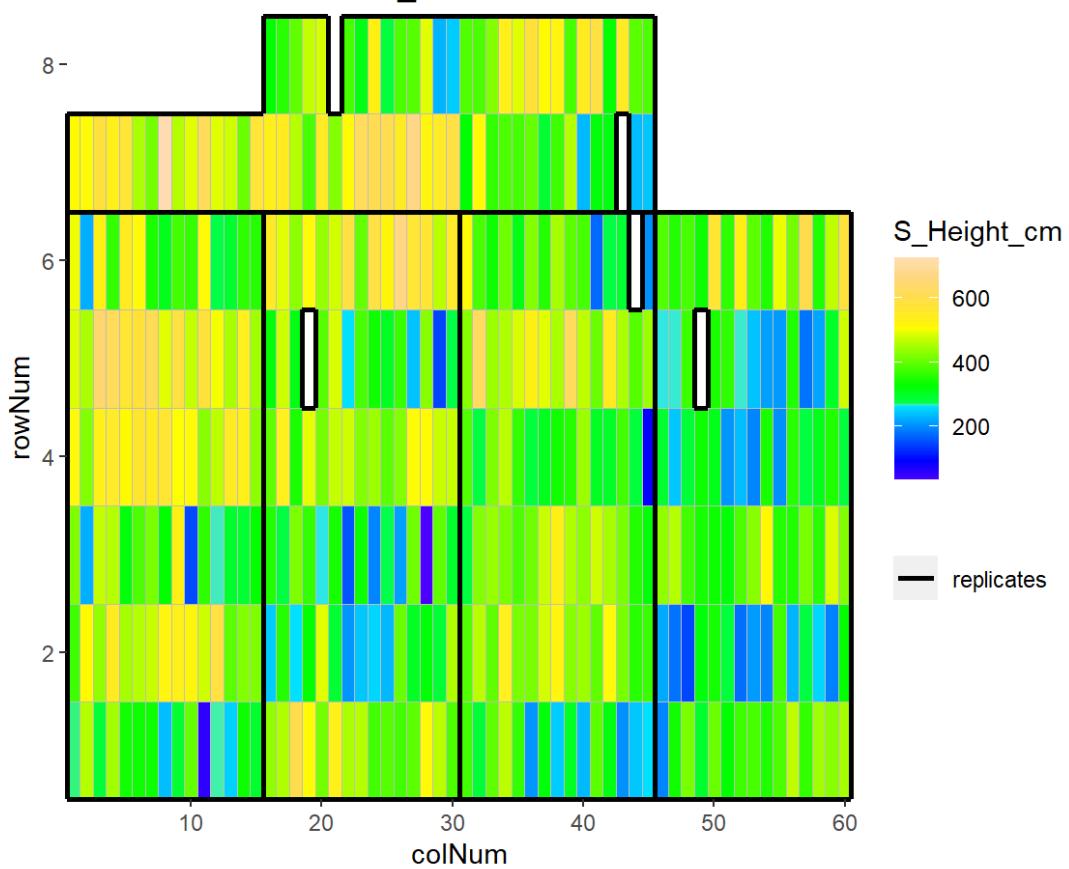
## EPPN2020\_M3P - 2020-03-01



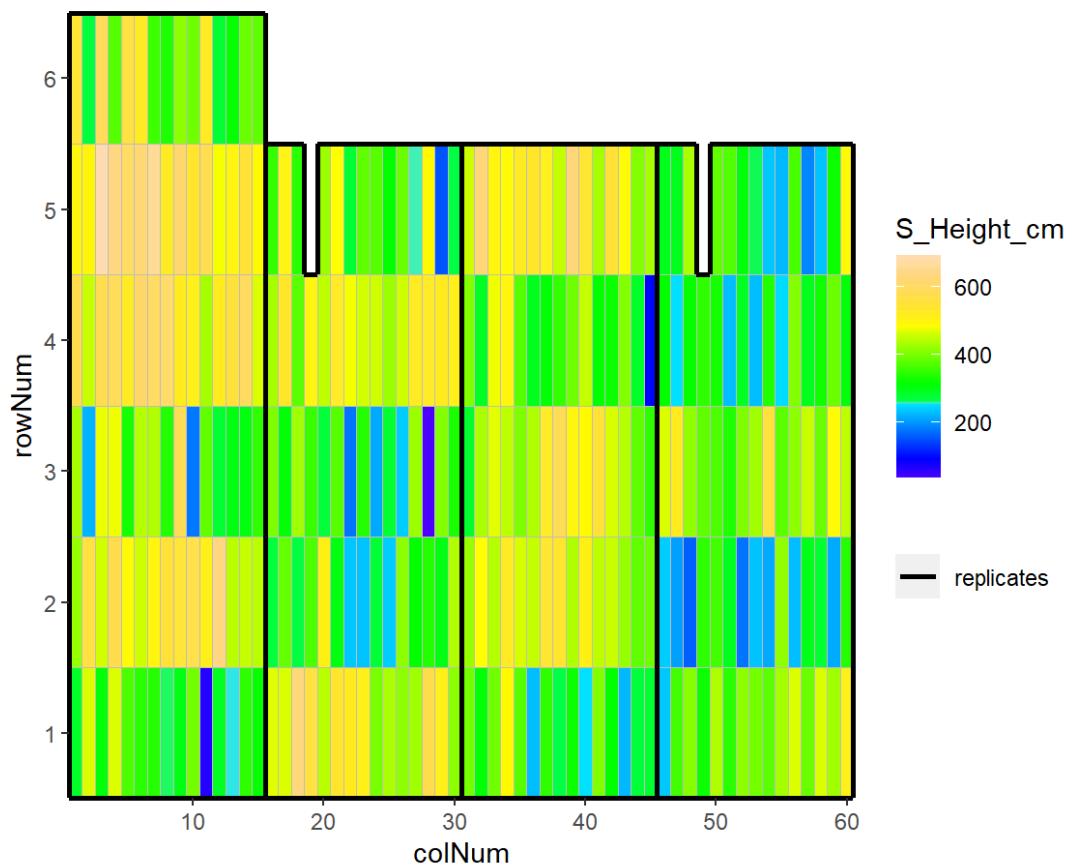
## EPPN2020\_M3P - 2020-03-02



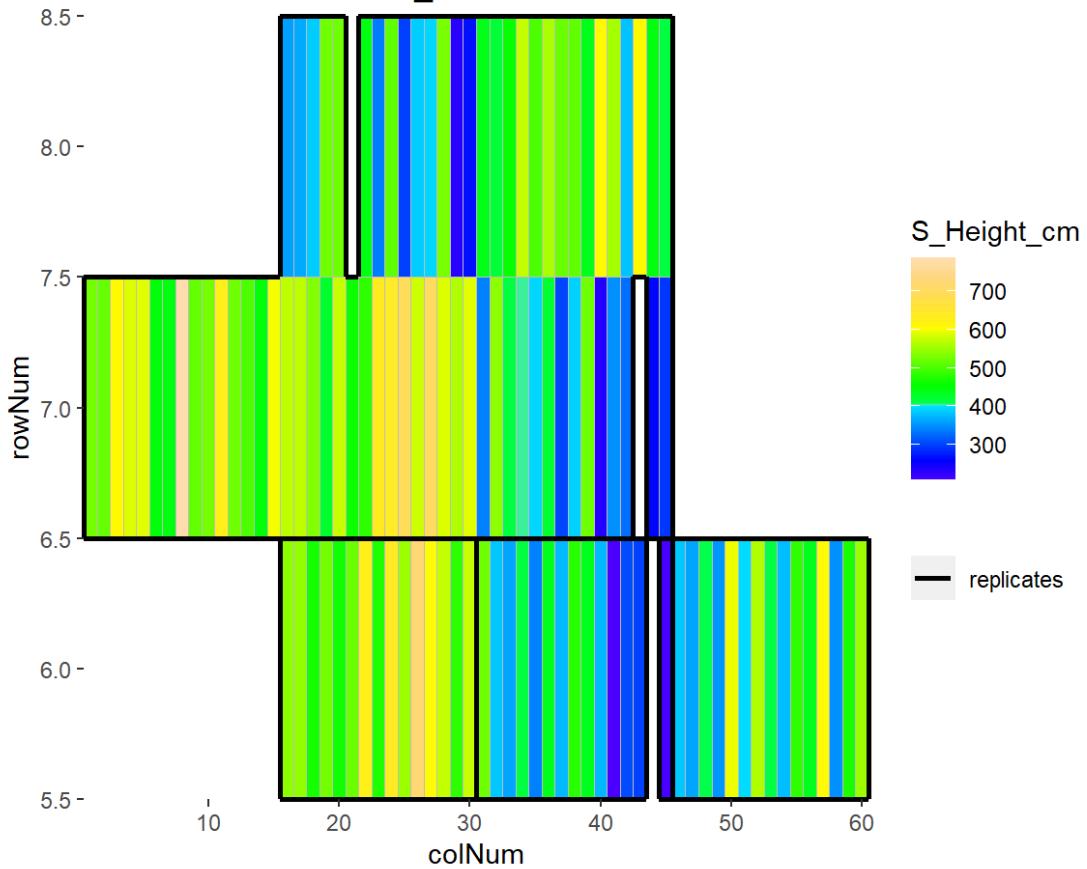
## EPPN2020\_M3P - 2020-03-03



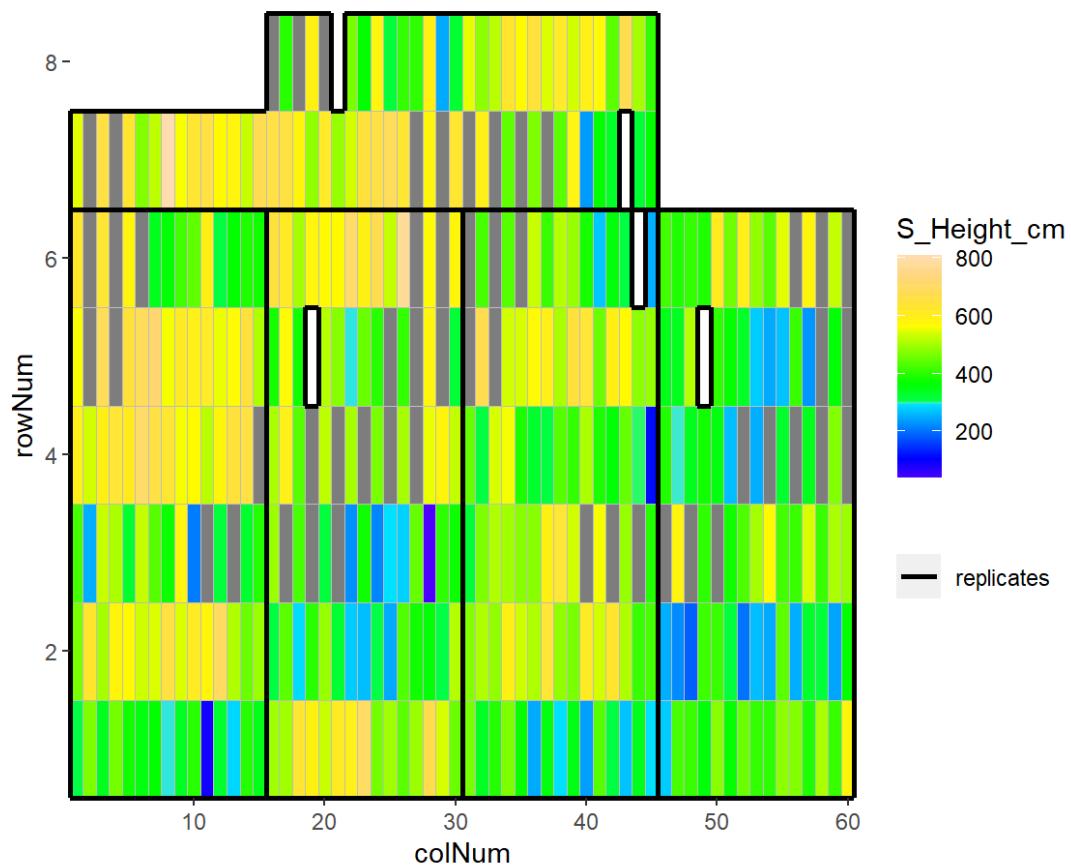
## EPPN2020\_M3P - 2020-03-04



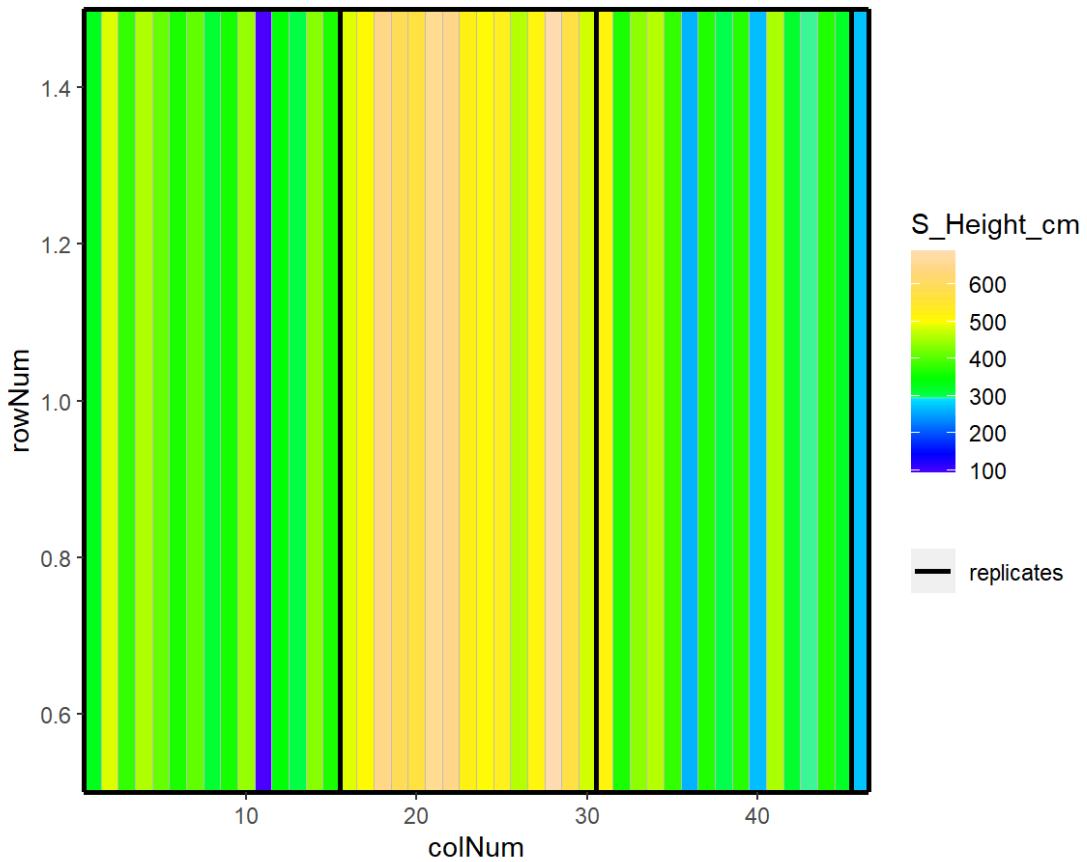
## EPPN2020\_M3P - 2020-03-05



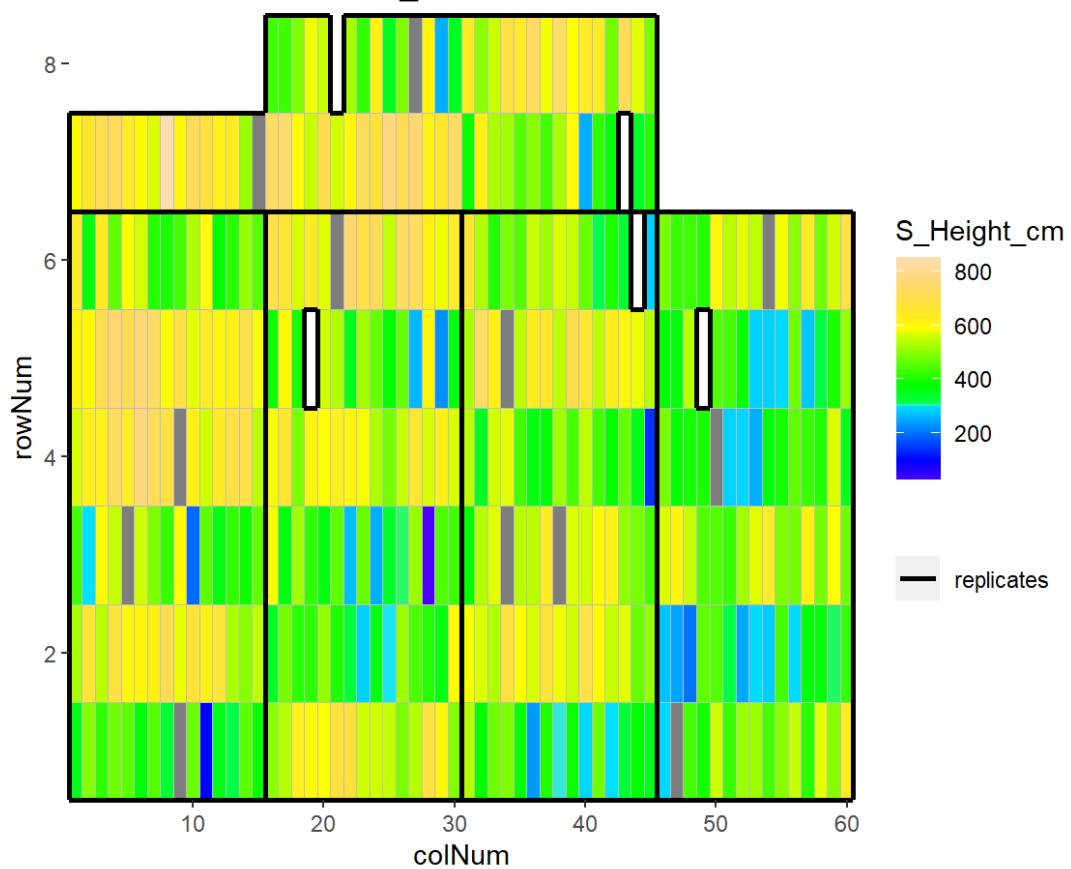
## EPPN2020\_M3P - 2020-03-06



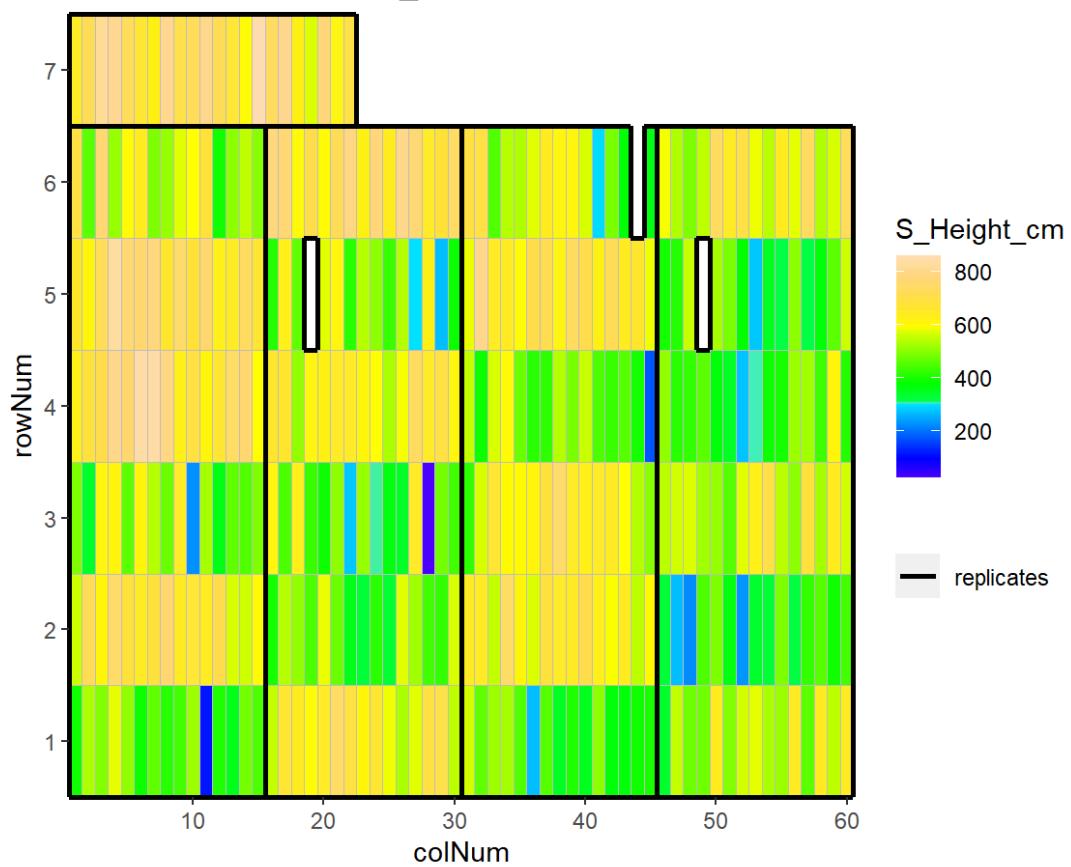
## EPPN2020\_M3P - 2020-03-07

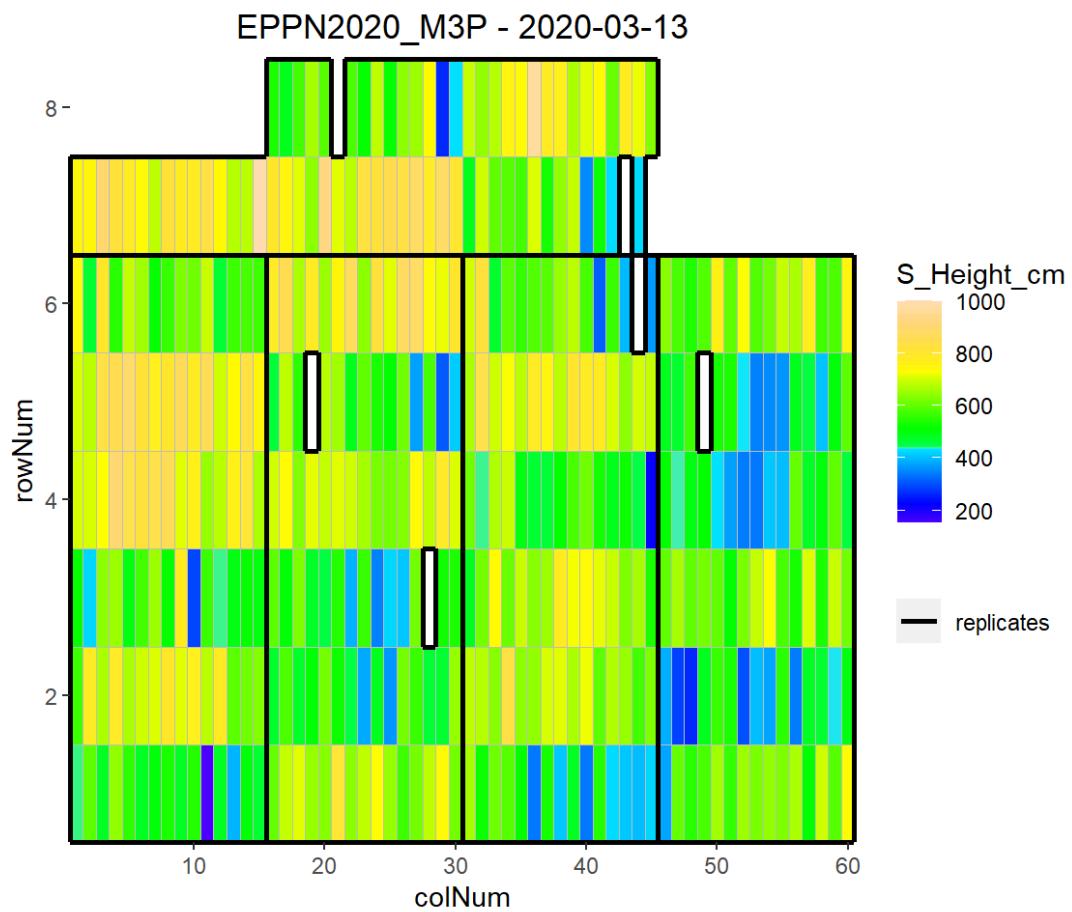
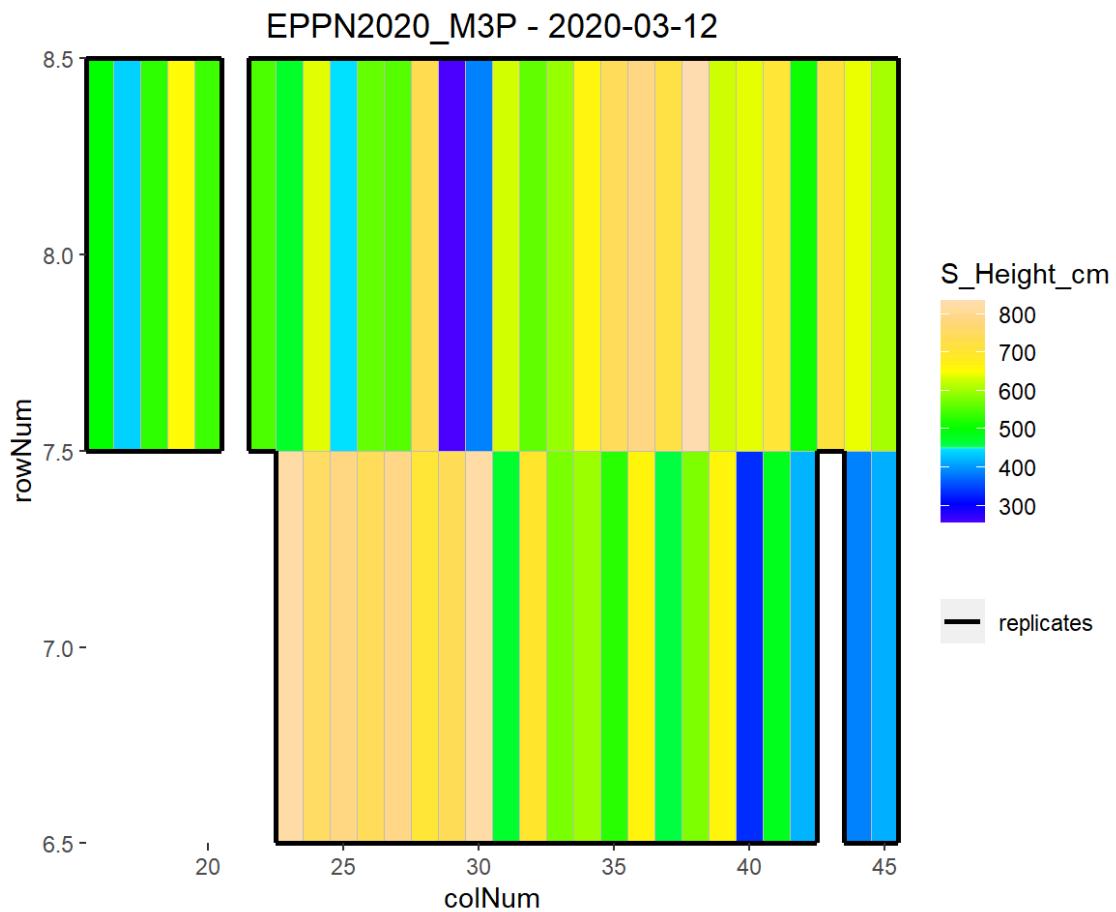


## EPPN2020\_M3P - 2020-03-08

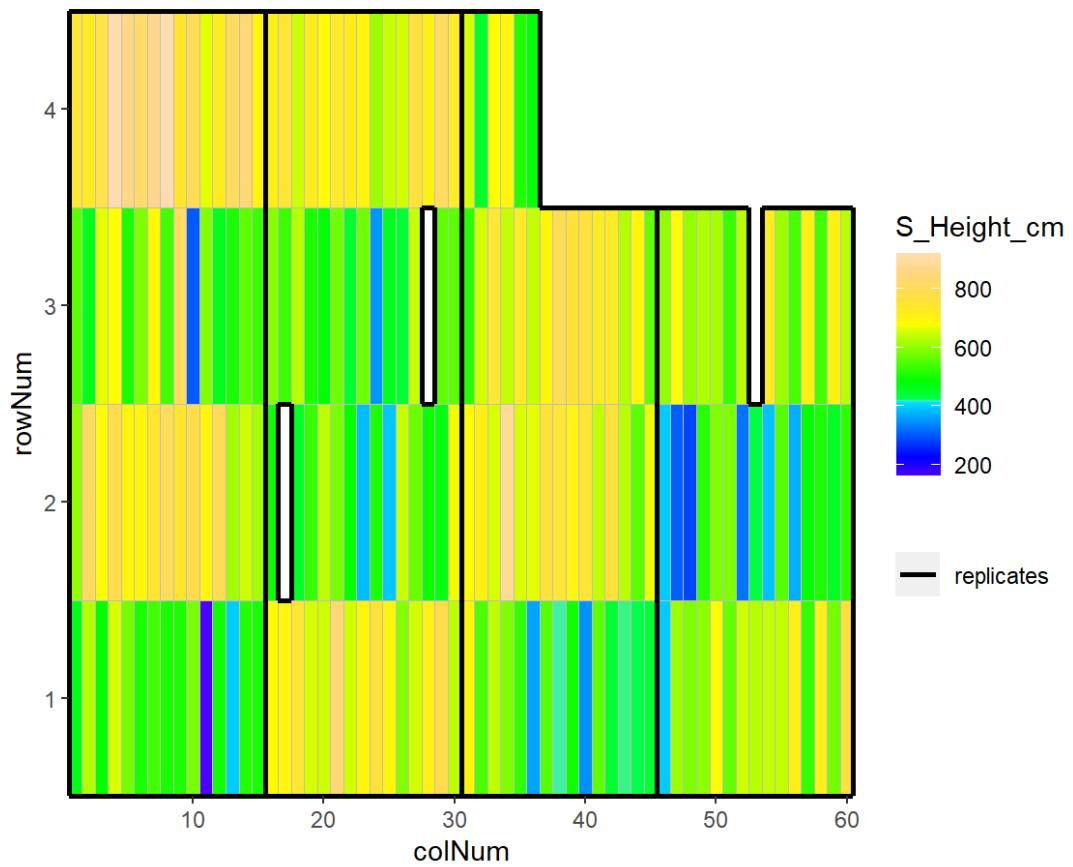


## EPPN2020\_M3P - 2020-03-11

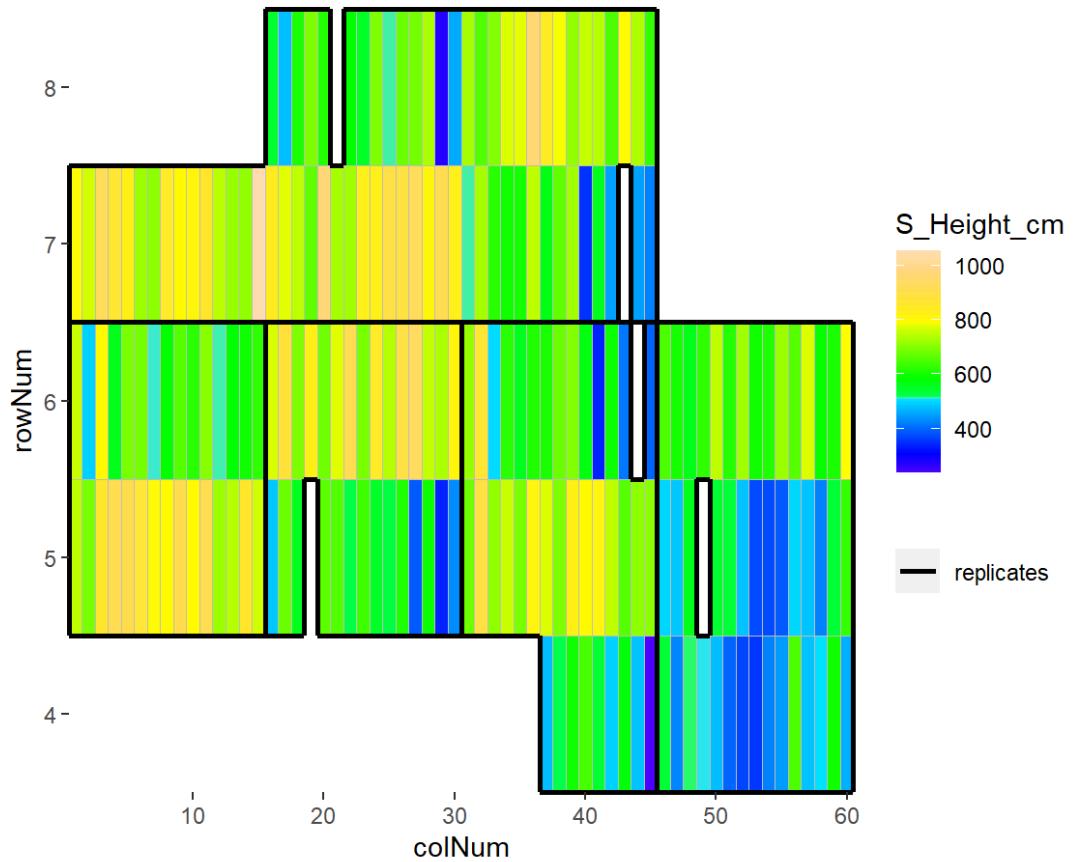




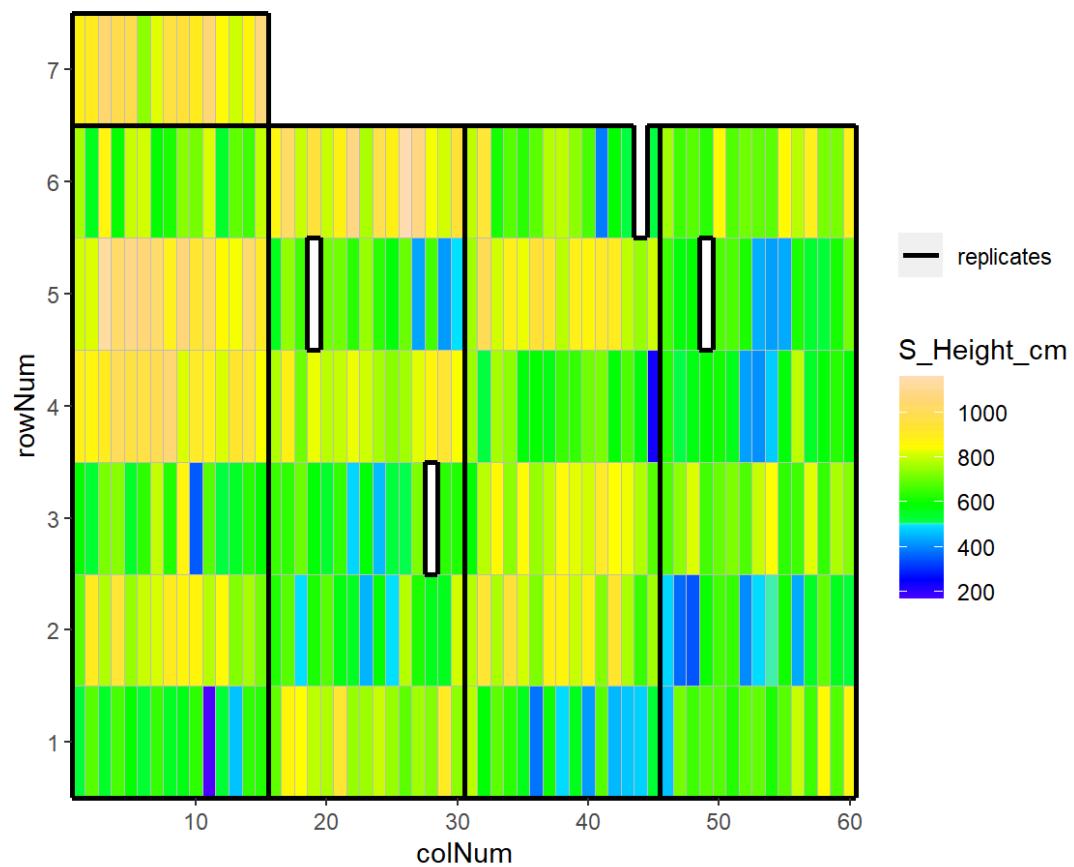
## EPPN2020\_M3P - 2020-03-14



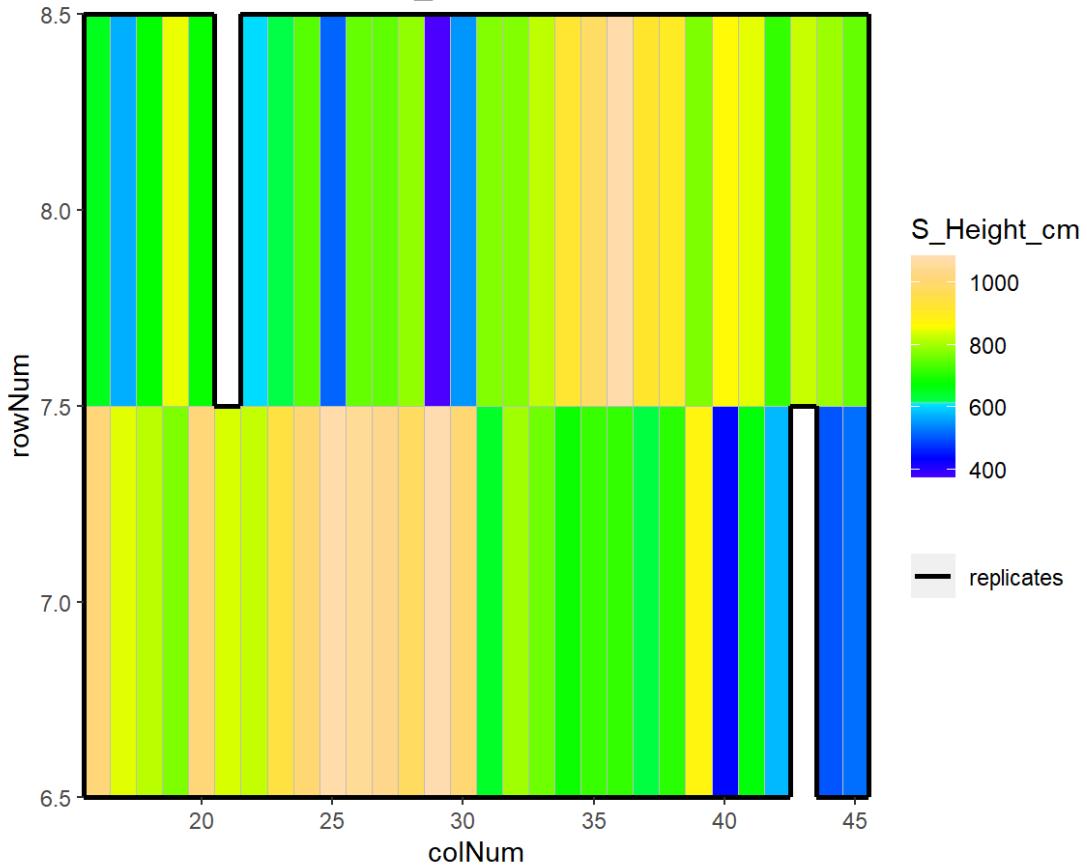
## EPPN2020\_M3P - 2020-03-15



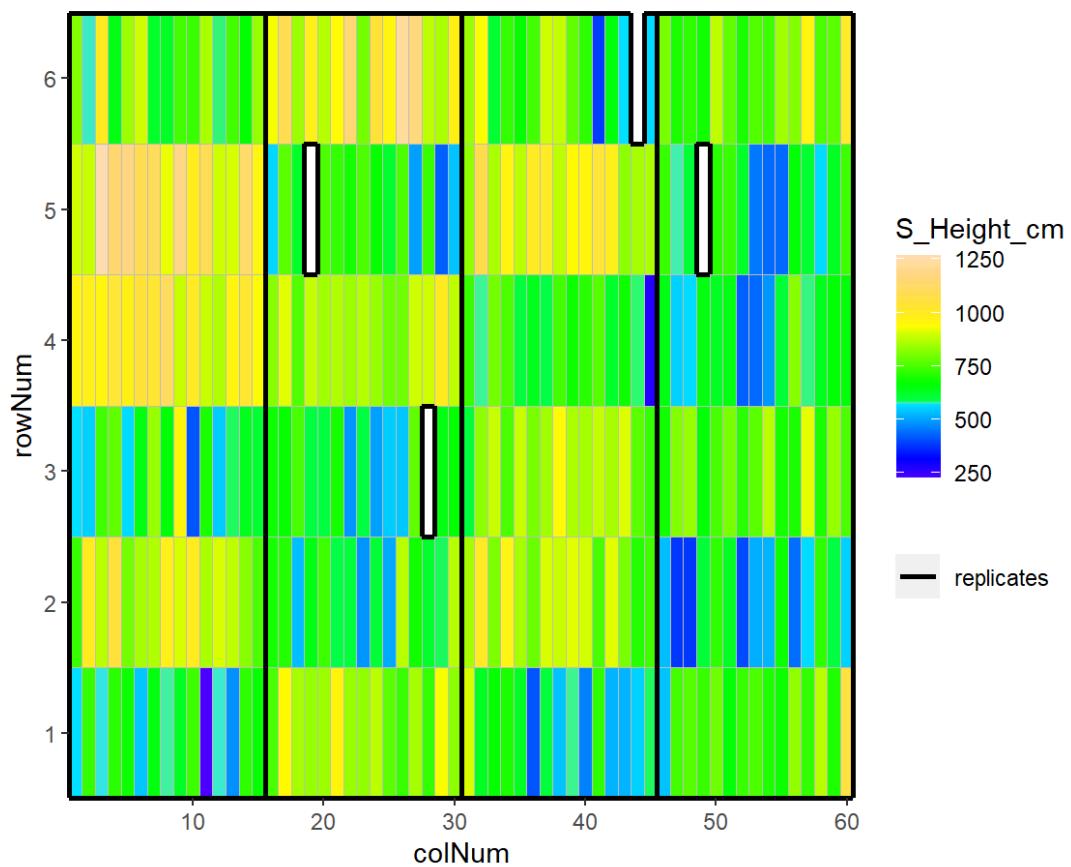
## EPPN2020\_M3P - 2020-03-18



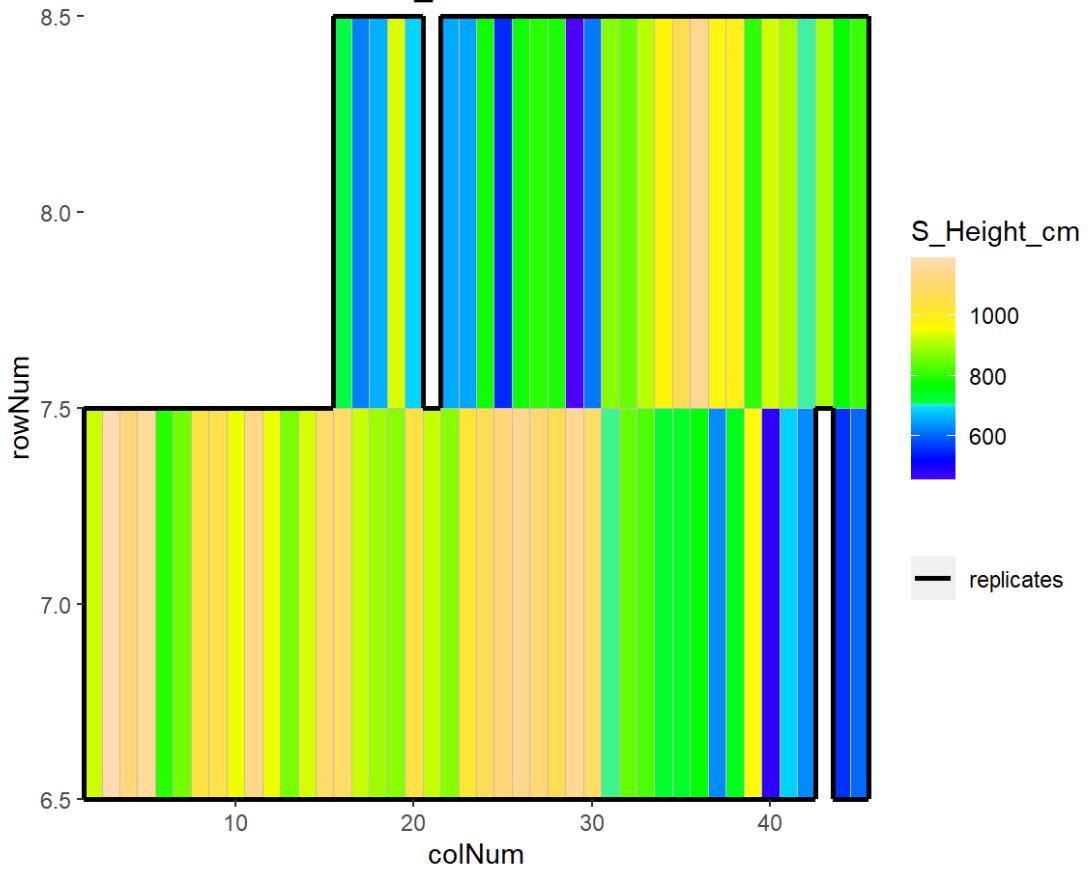
## EPPN2020\_M3P - 2020-03-19



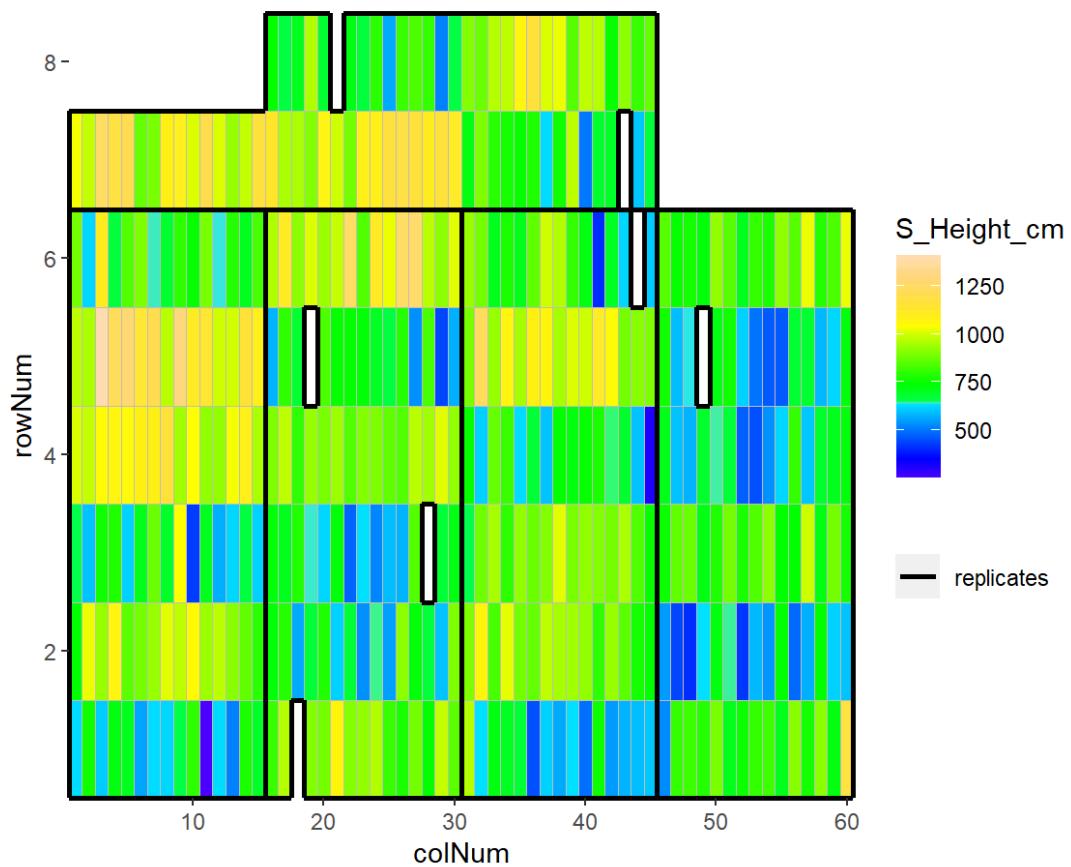
## EPPN2020\_M3P - 2020-03-20



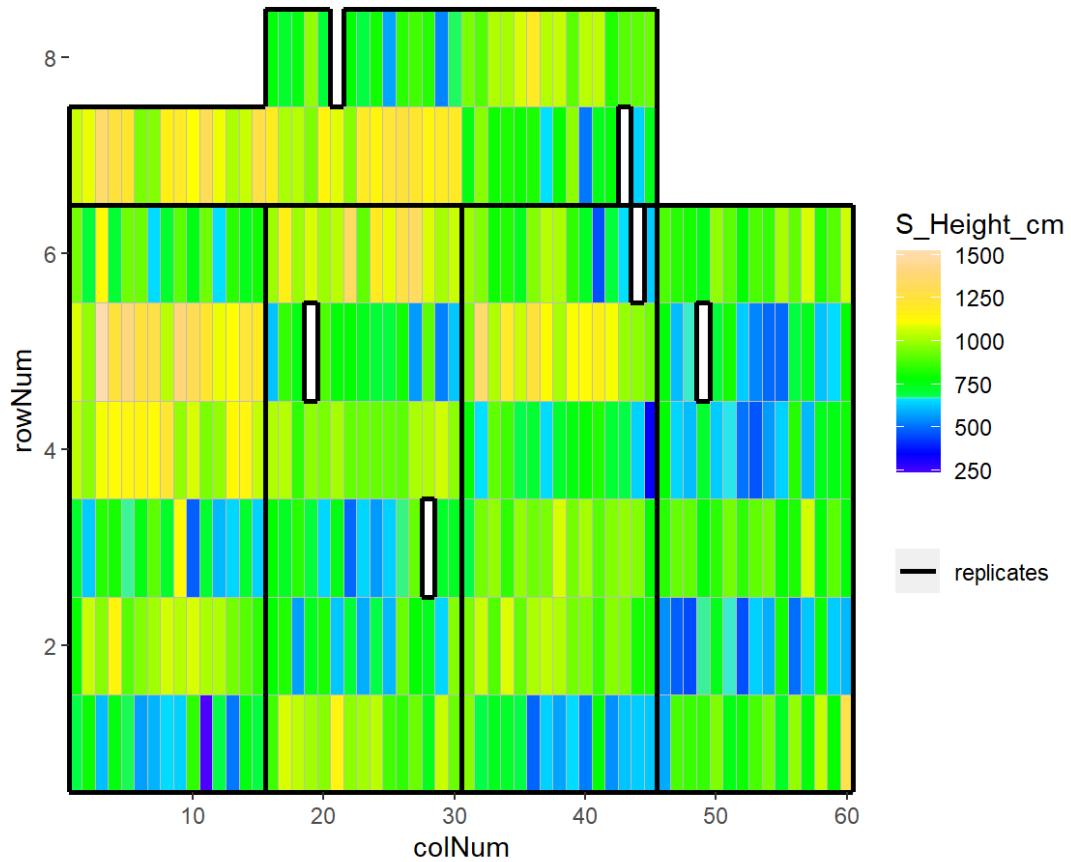
## EPPN2020\_M3P - 2020-03-21



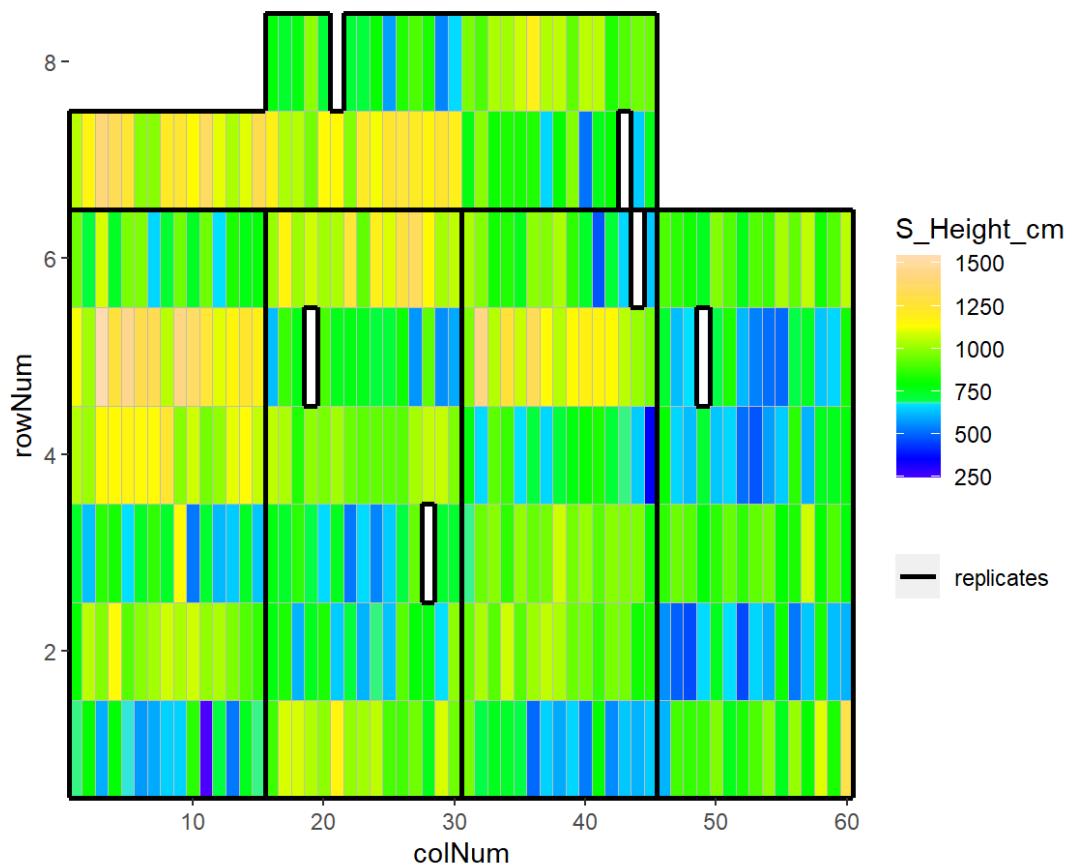
## EPPN2020\_M3P - 2020-03-22



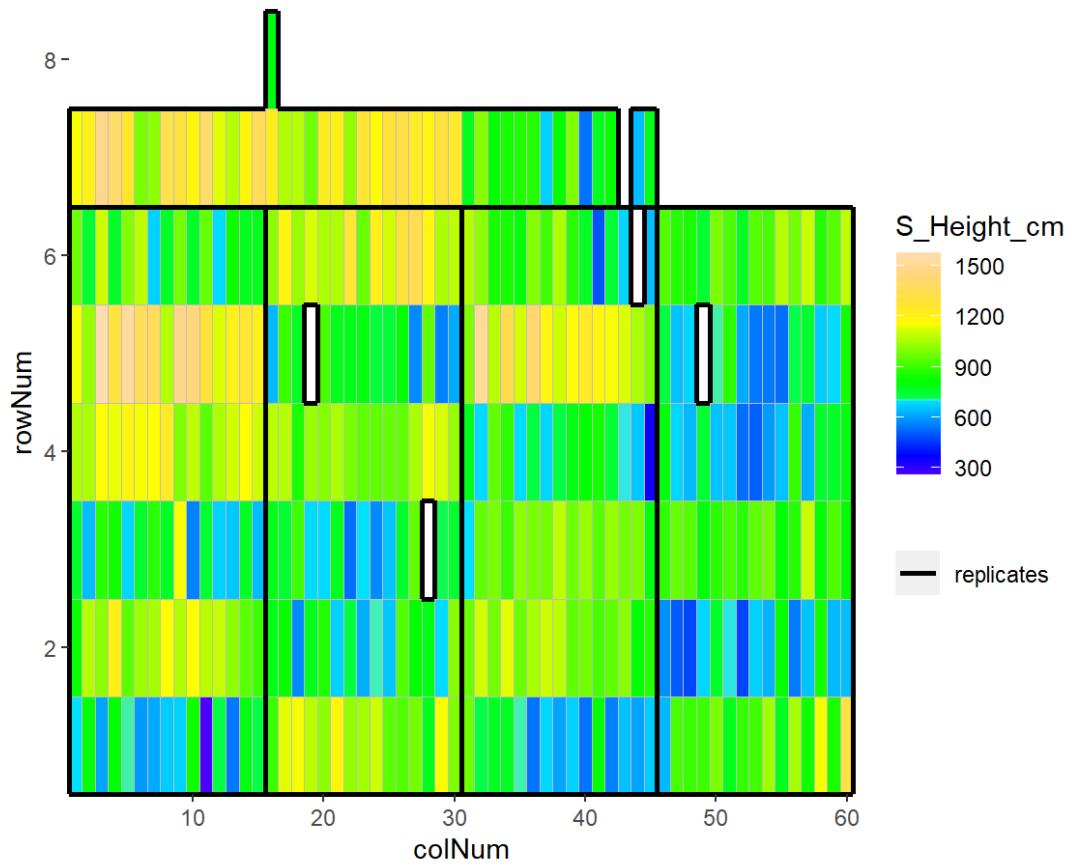
## EPPN2020\_M3P - 2020-03-24



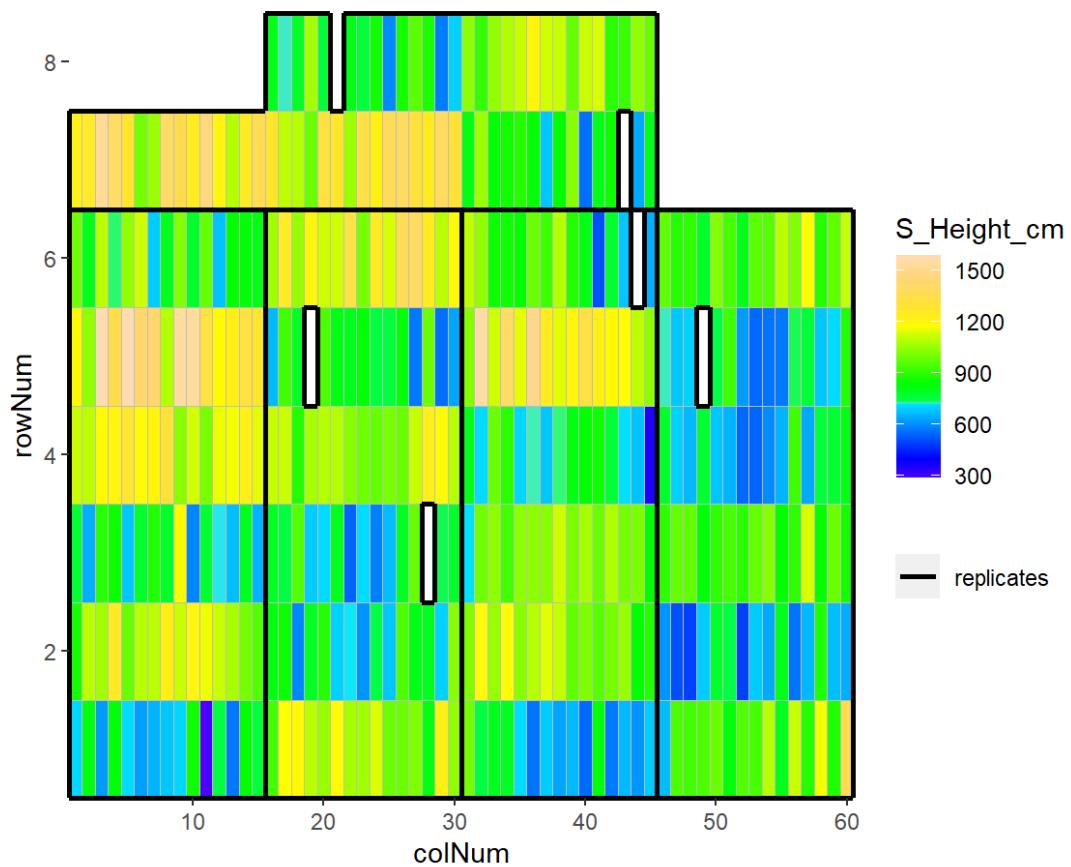
## EPPN2020\_M3P - 2020-03-25



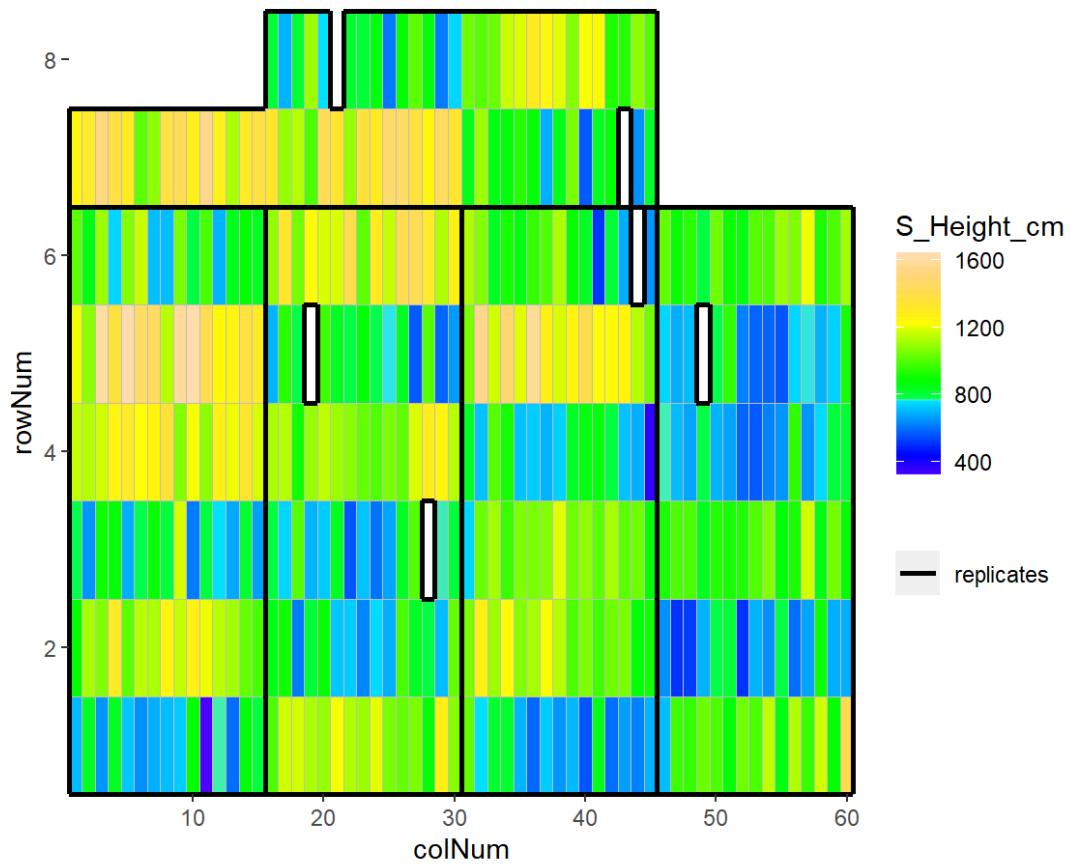
## EPPN2020\_M3P - 2020-03-26



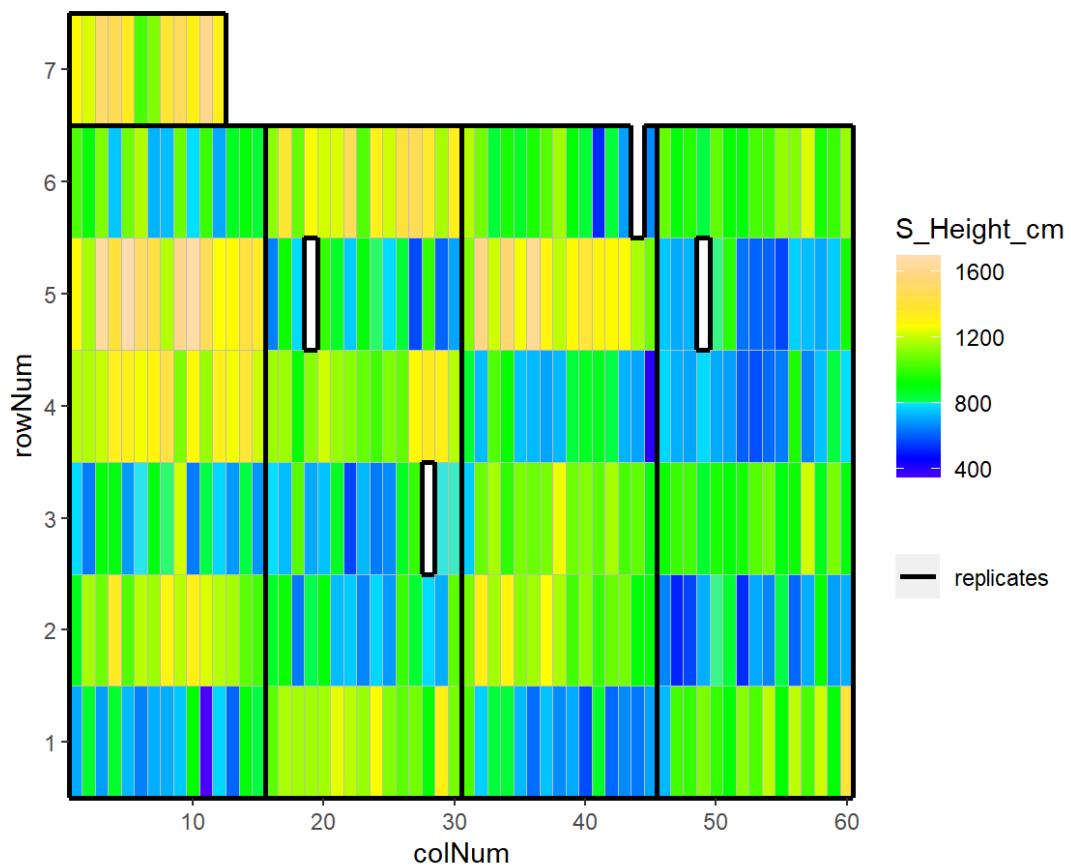
## EPPN2020\_M3P - 2020-03-27



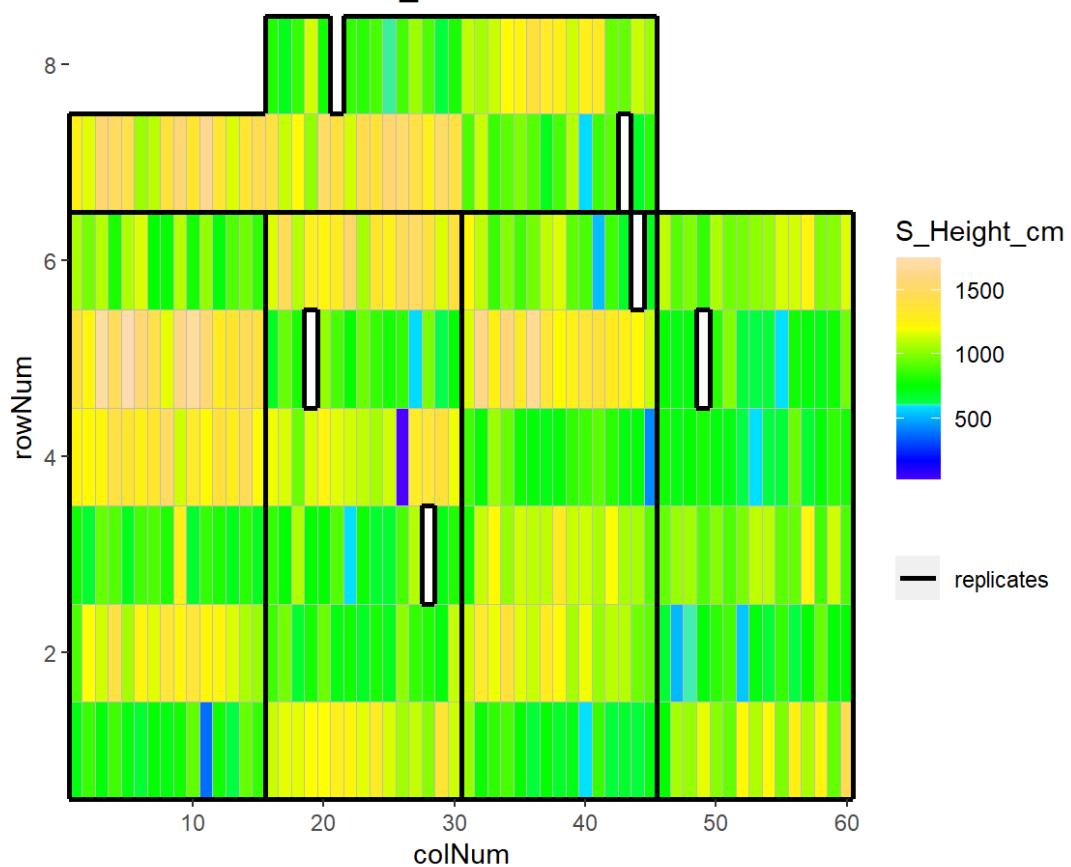
## EPPN2020\_M3P - 2020-03-28



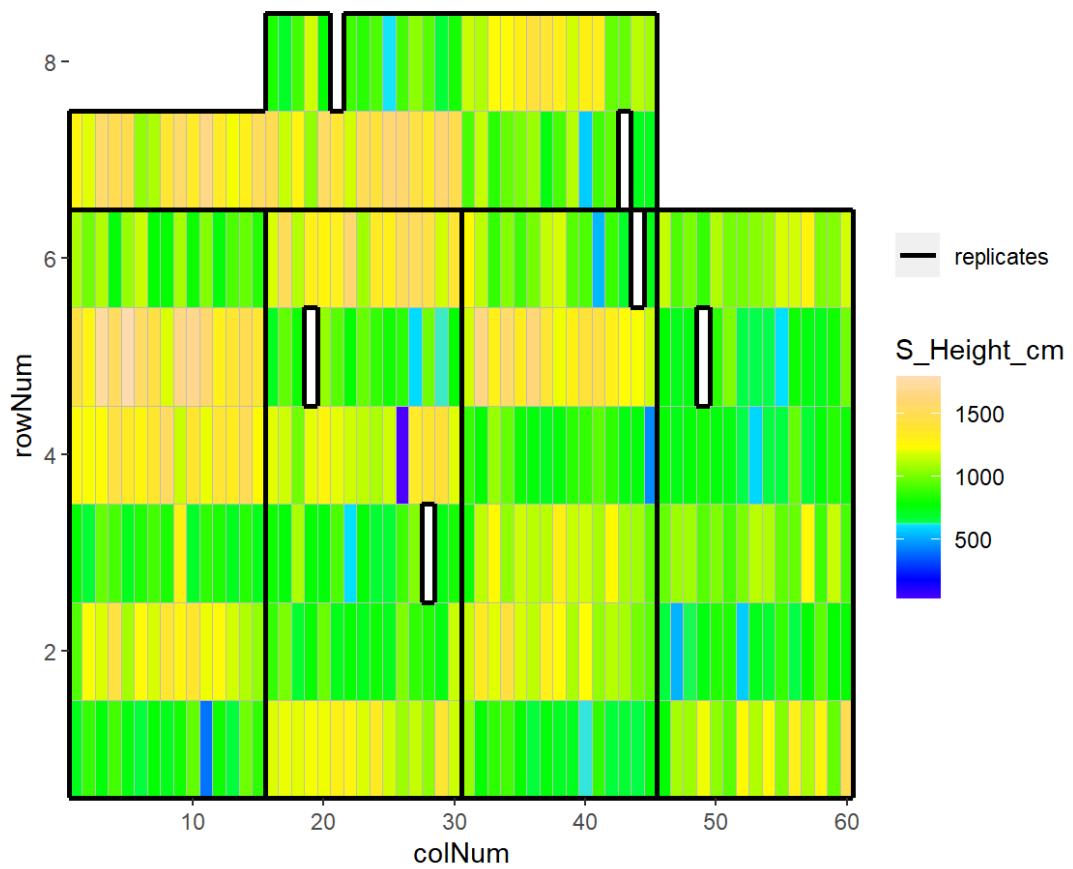
## EPPN2020\_M3P - 2020-03-29



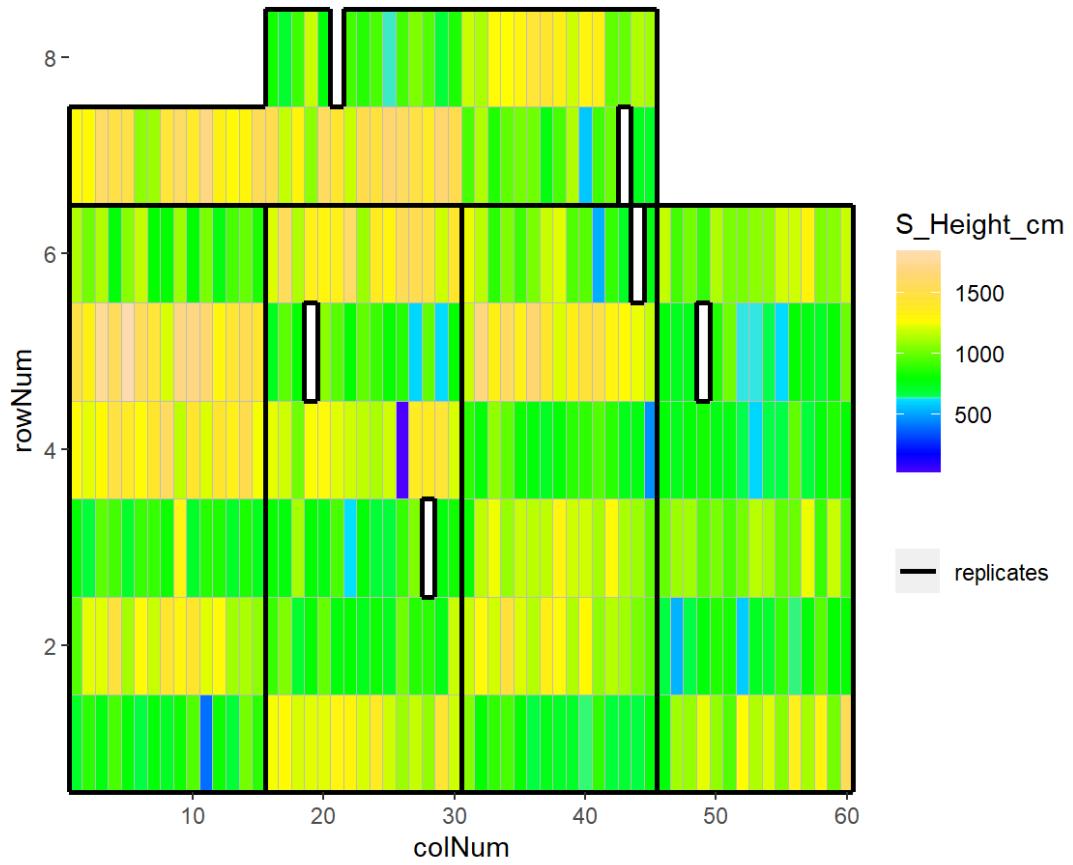
## EPPN2020\_M3P - 2020-03-30



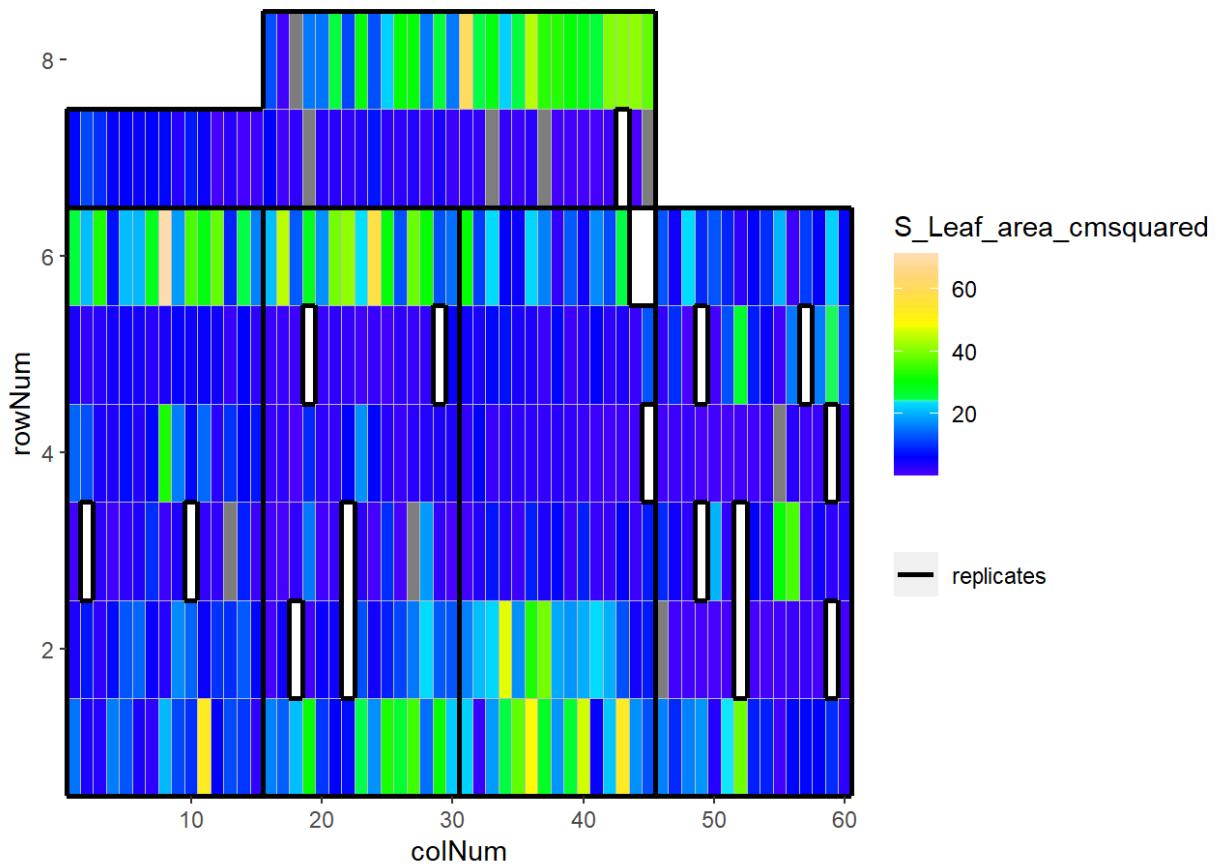
## EPPN2020\_M3P - 2020-03-31



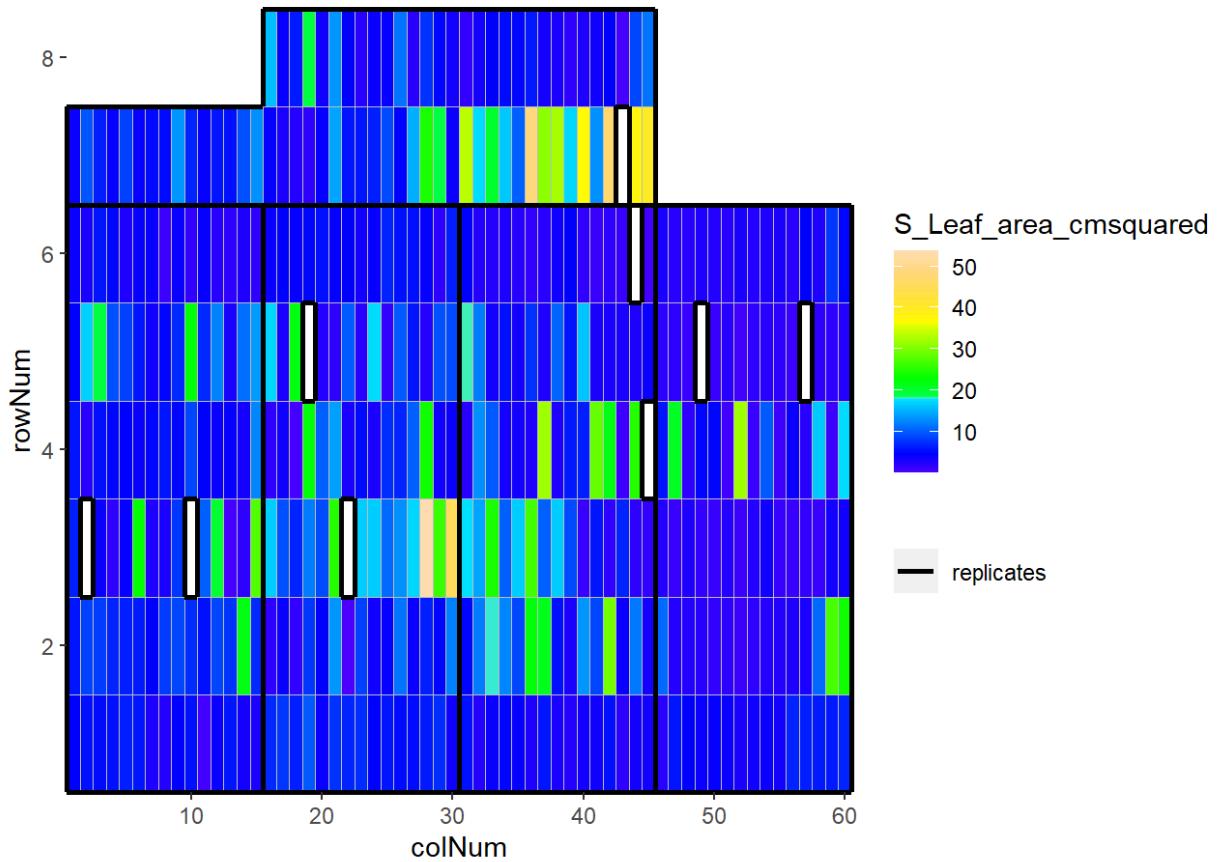
## EPPN2020\_M3P - 2020-04-01



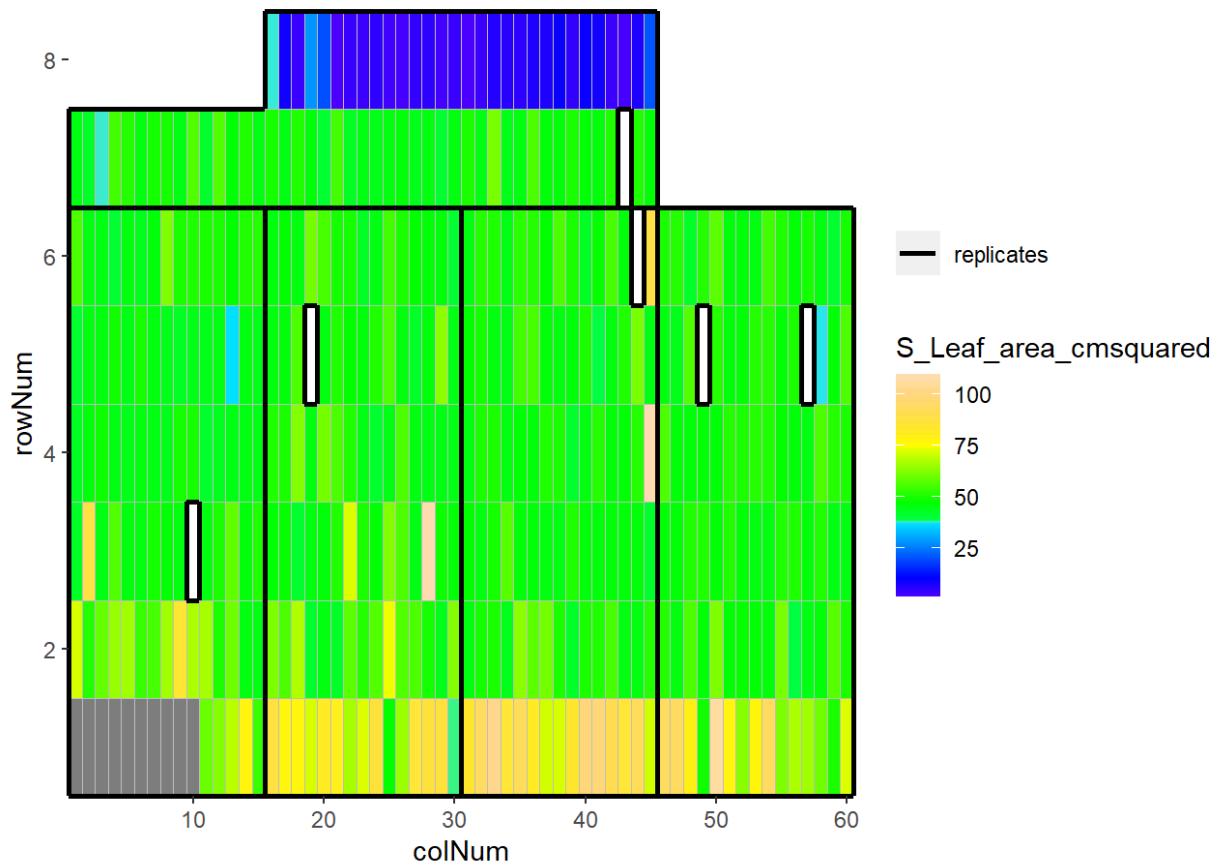
## EPPN2020\_M3P - 2020-02-14



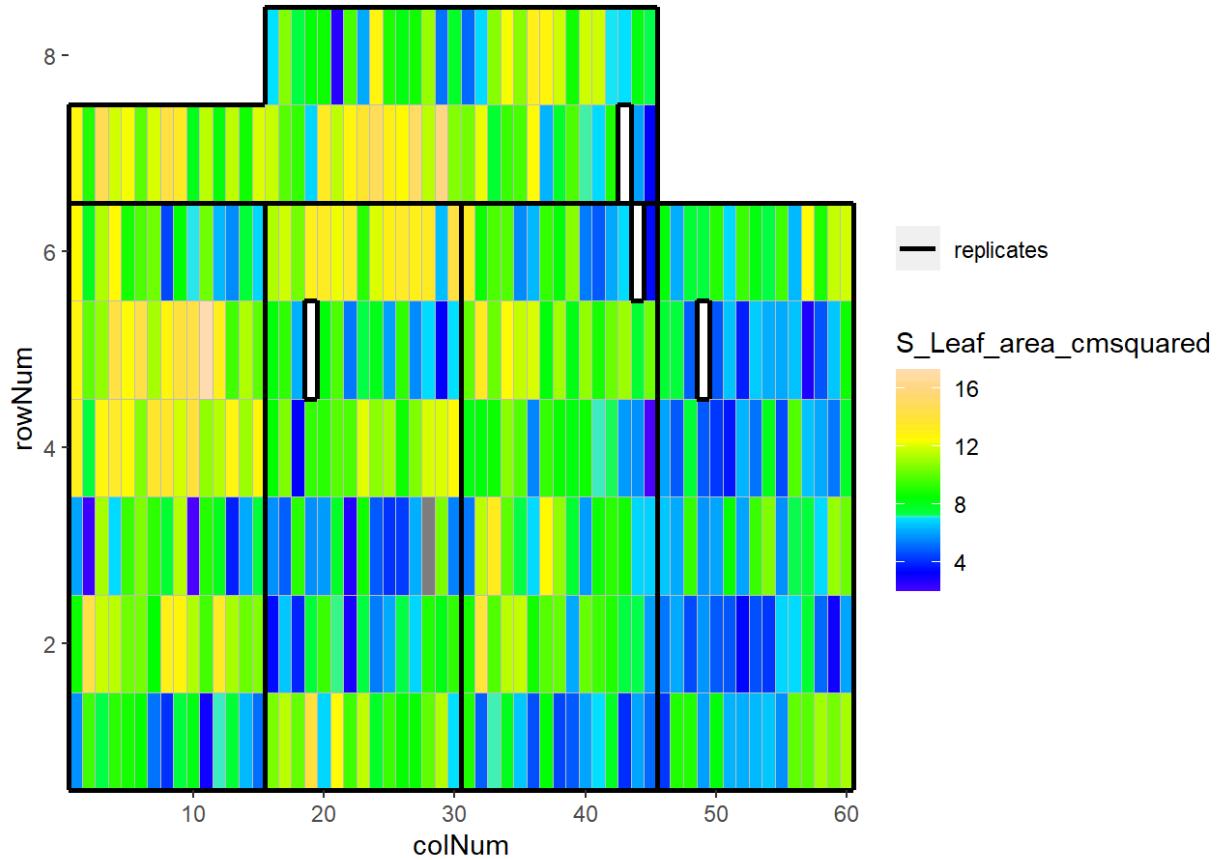
## EPPN2020\_M3P - 2020-02-15



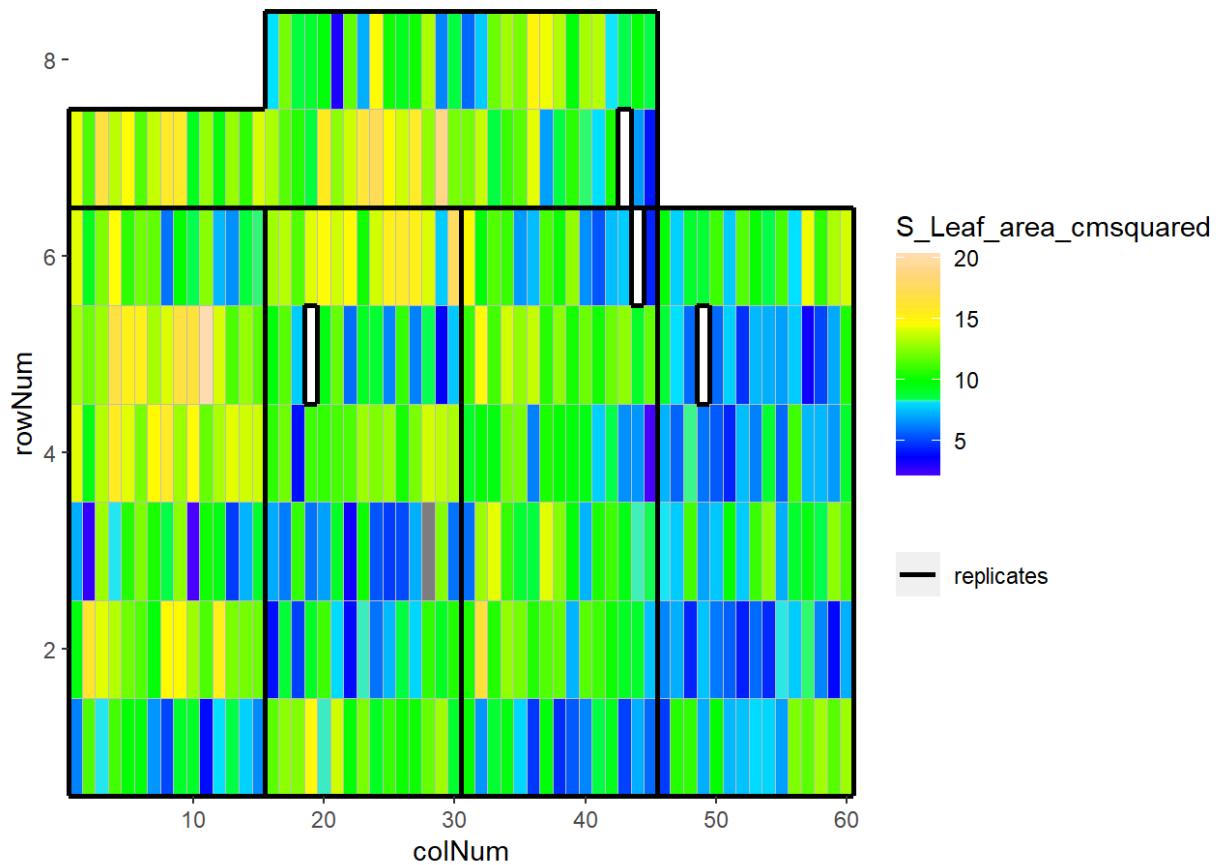
## EPPN2020\_M3P - 2020-02-16



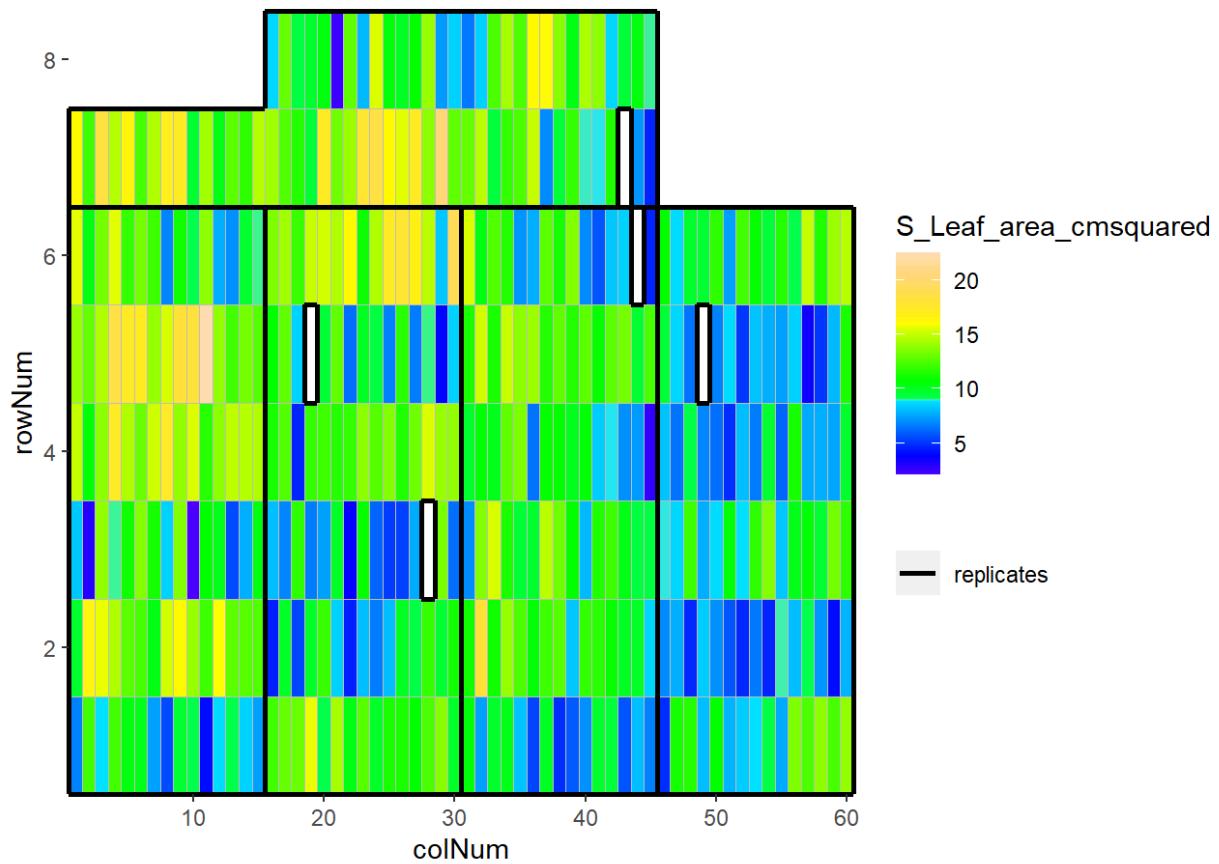
## EPPN2020\_M3P - 2020-02-17



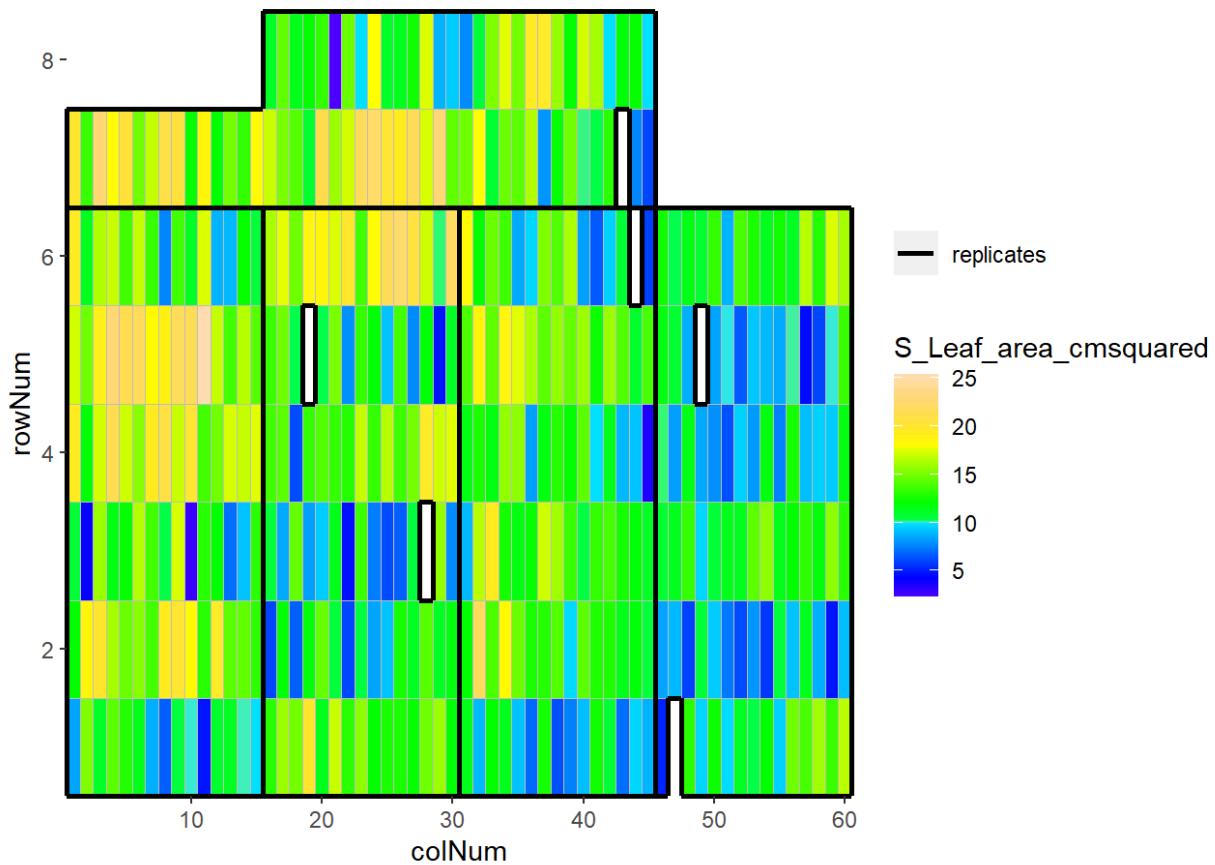
## EPPN2020\_M3P - 2020-02-18



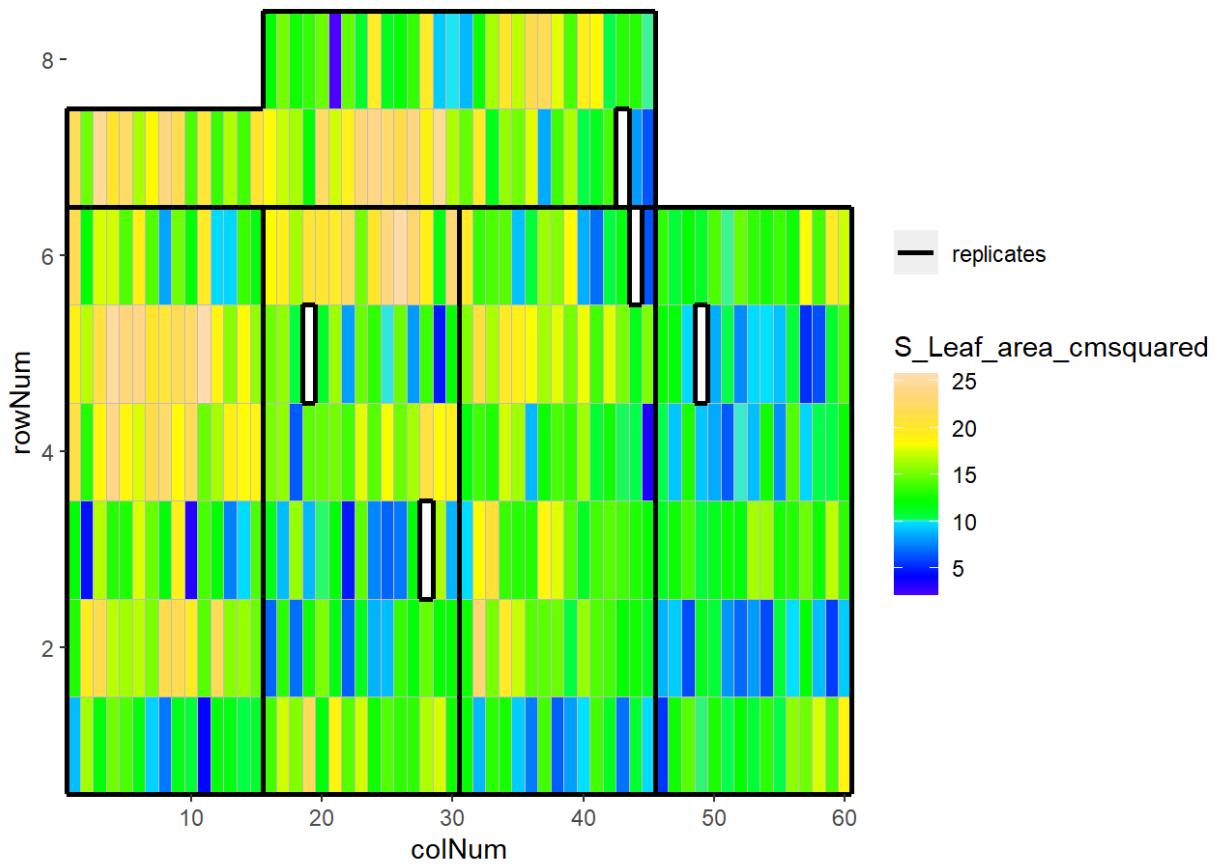
## EPPN2020\_M3P - 2020-02-19



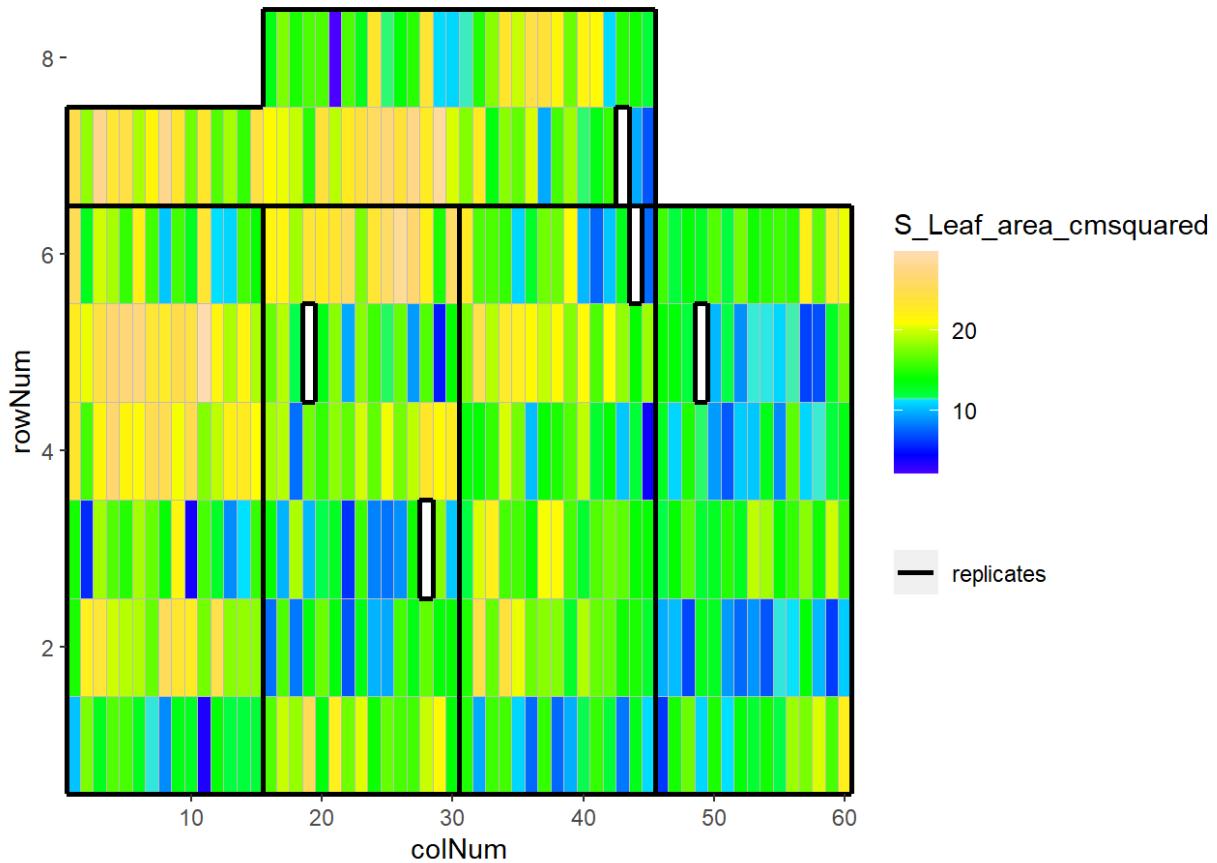
## EPPN2020\_M3P - 2020-02-20



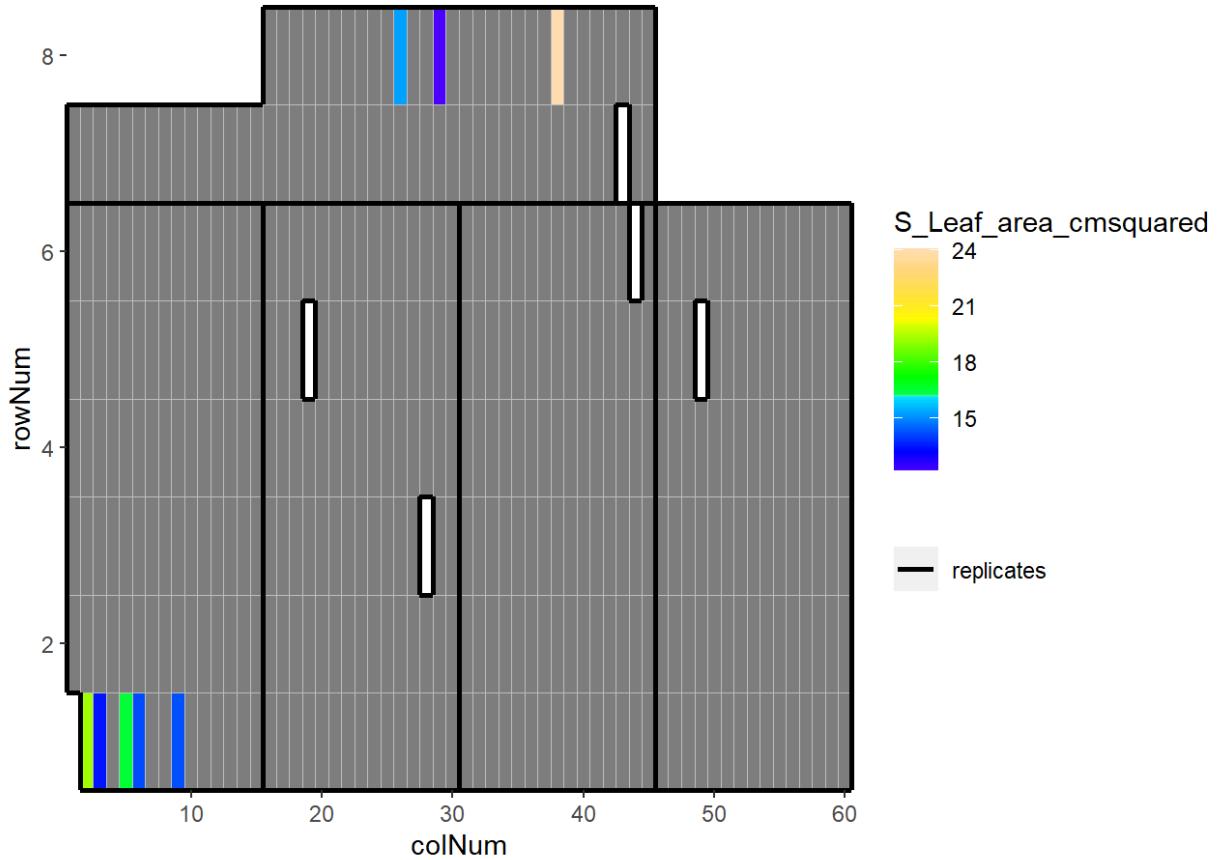
## EPPN2020\_M3P - 2020-02-21



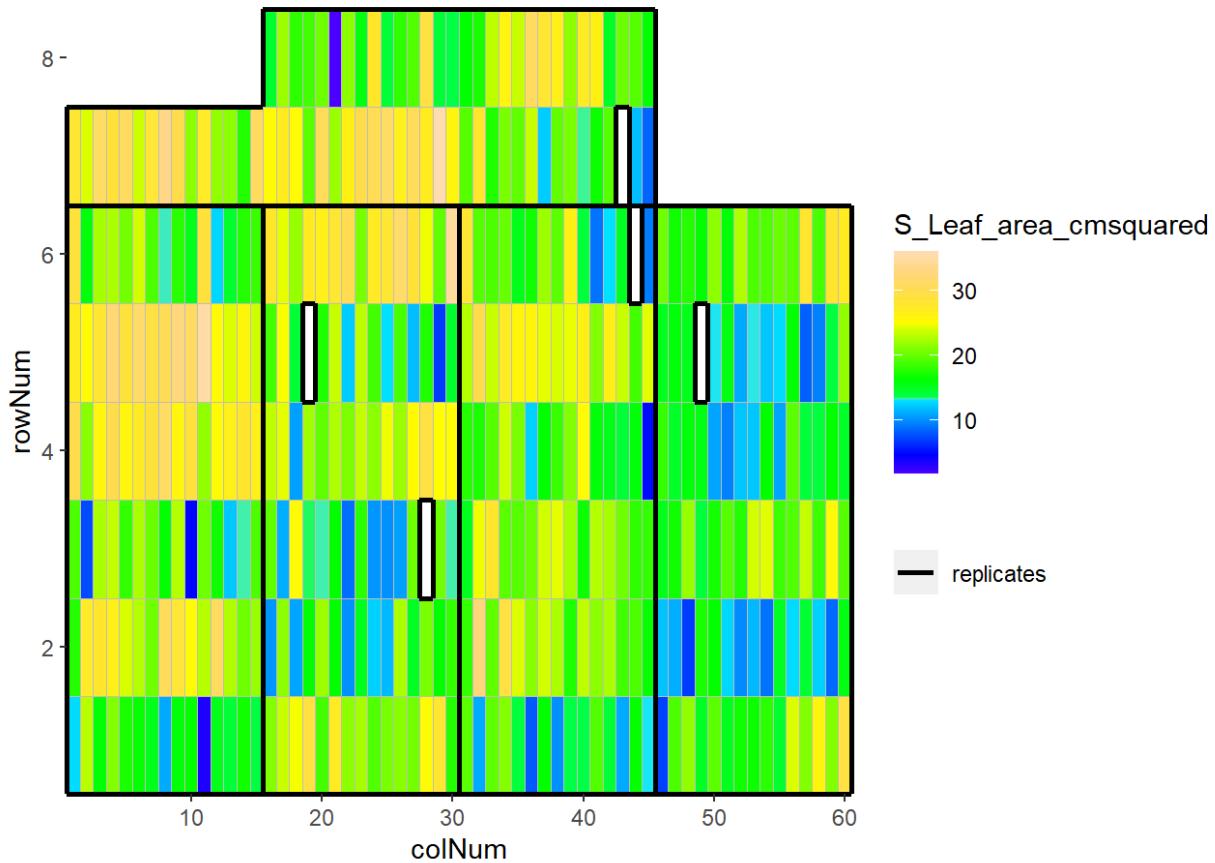
## EPPN2020\_M3P - 2020-02-22



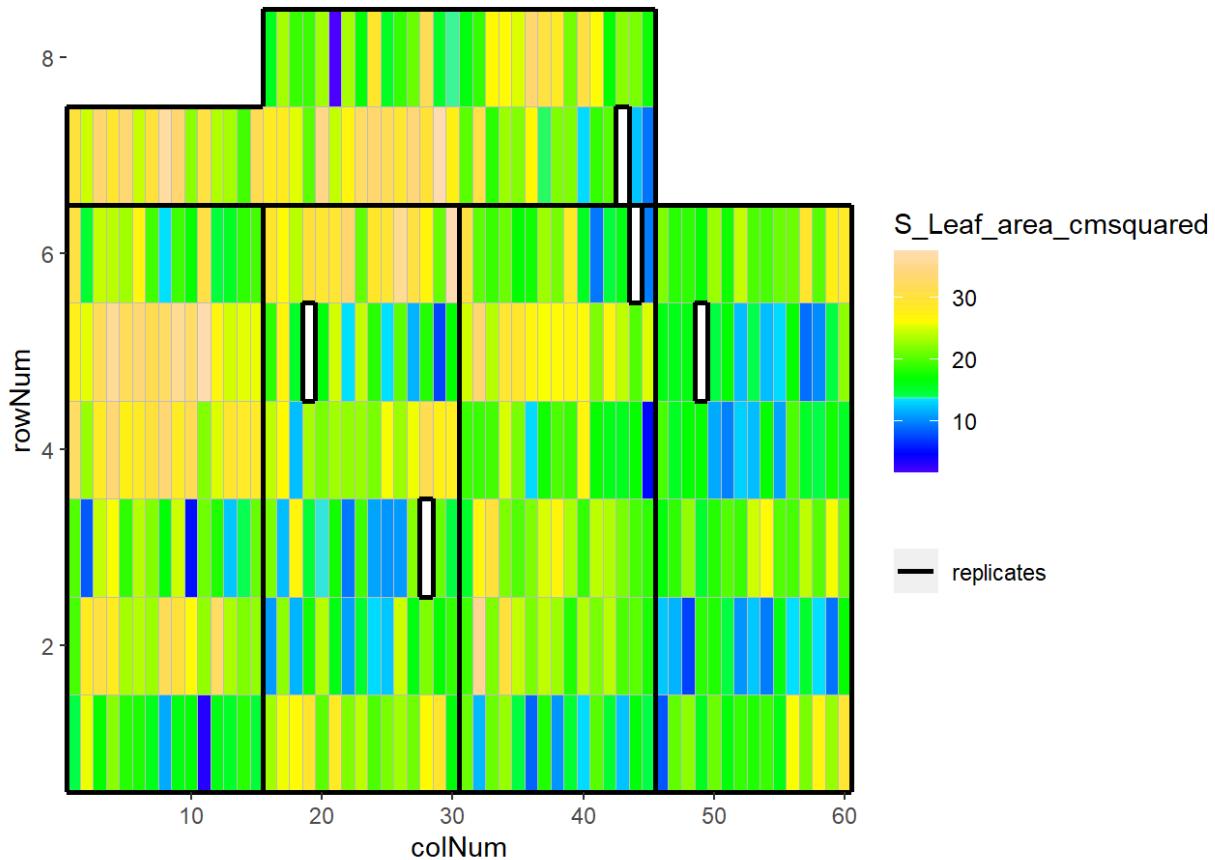
## EPPN2020\_M3P - 2020-02-23



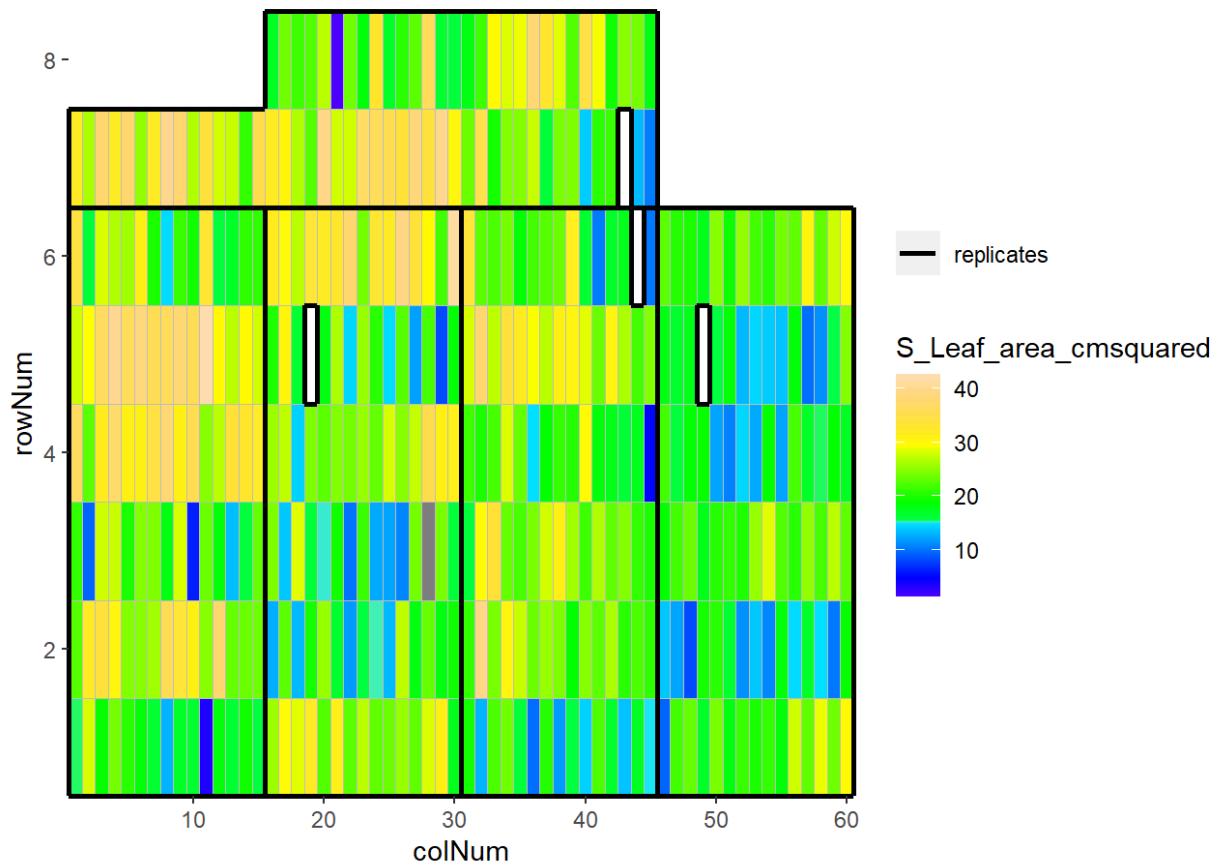
## EPPN2020\_M3P - 2020-02-24



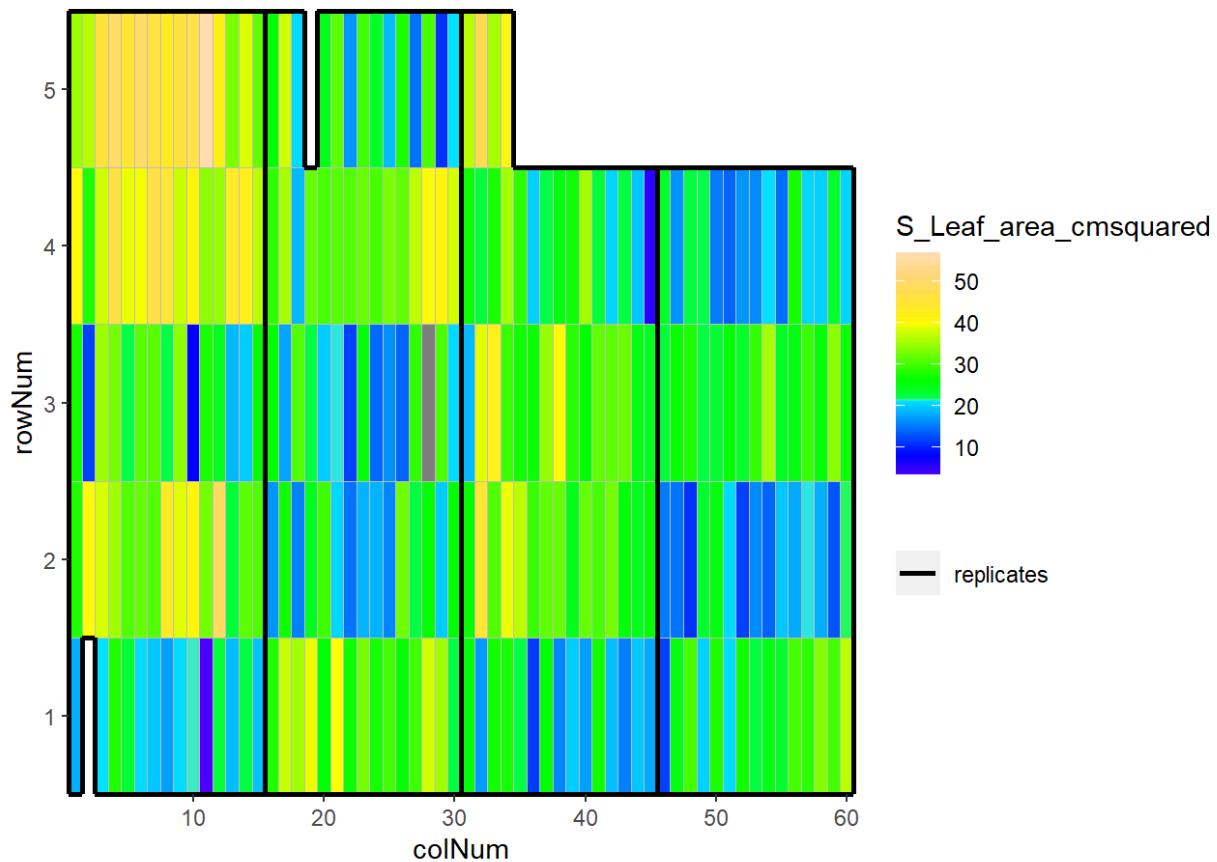
## EPPN2020\_M3P - 2020-02-25



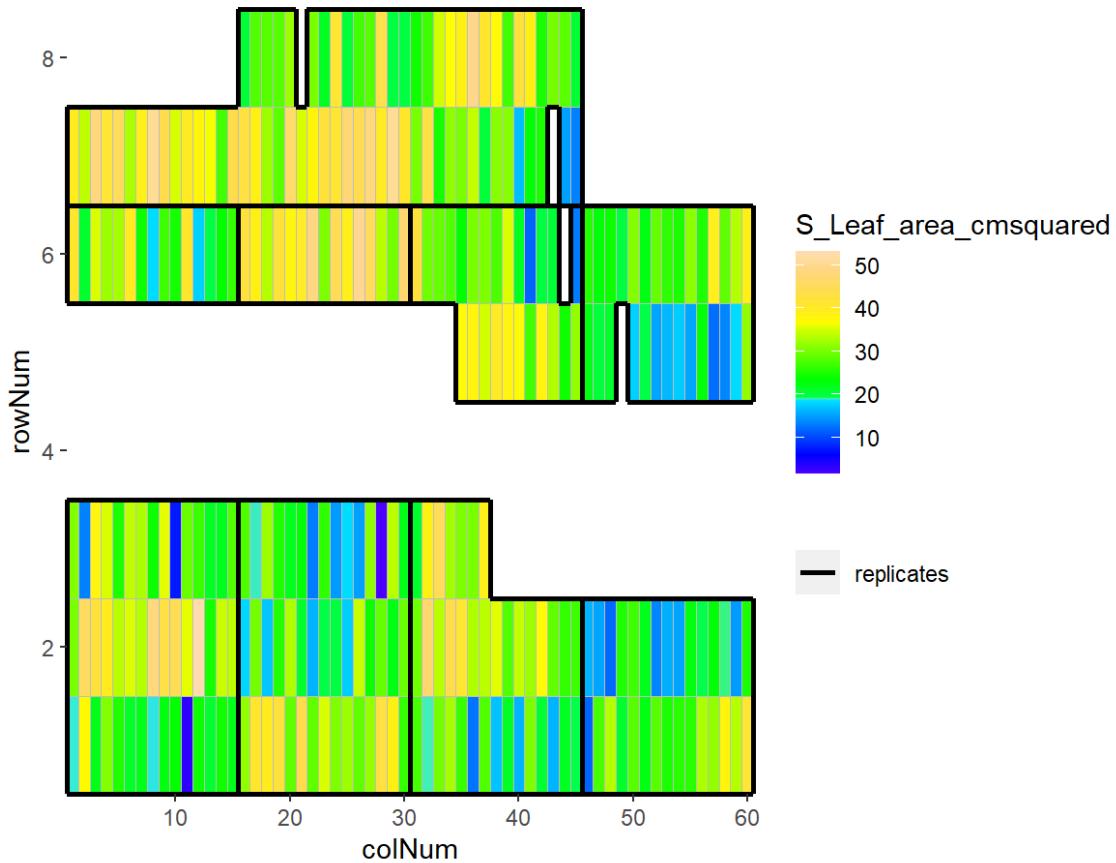
## EPPN2020\_M3P - 2020-02-26



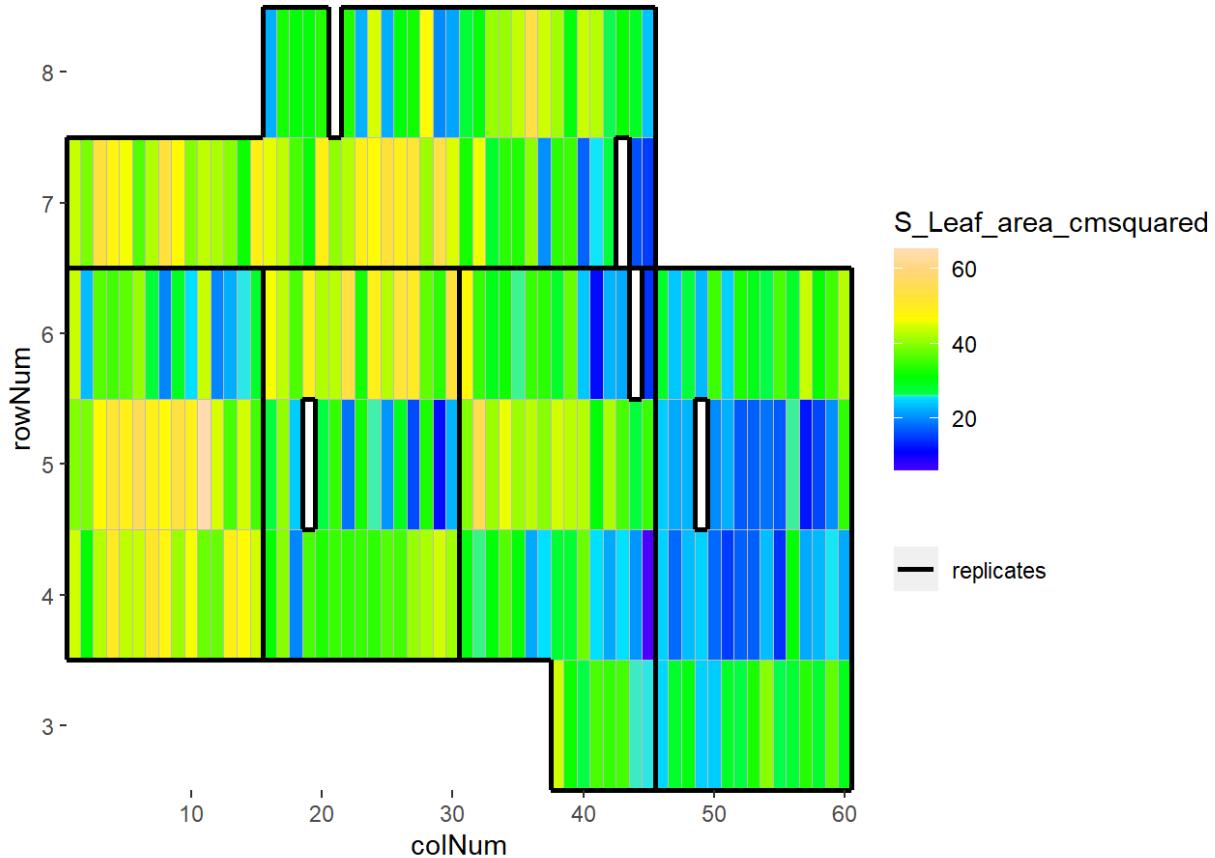
## EPPN2020\_M3P - 2020-02-28



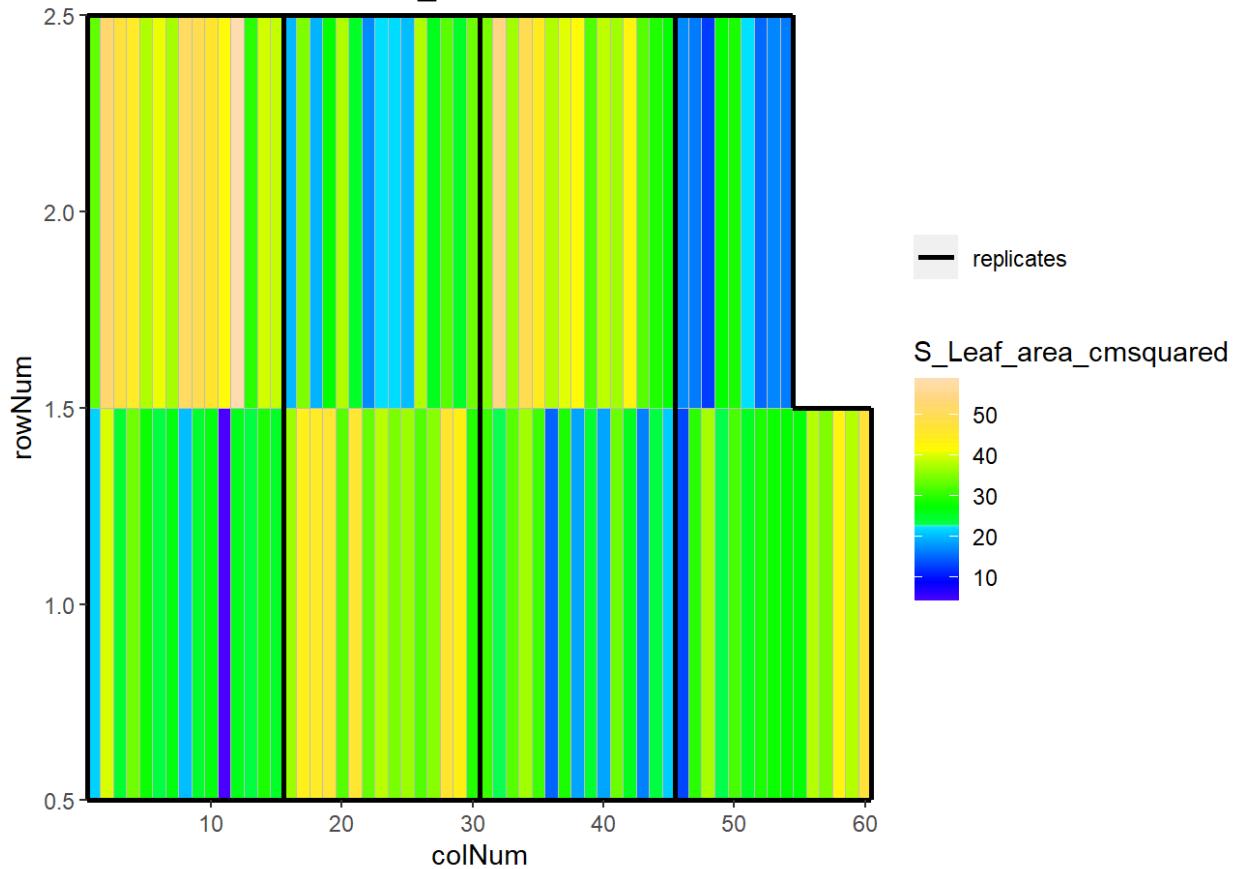
## EPPN2020\_M3P - 2020-02-29



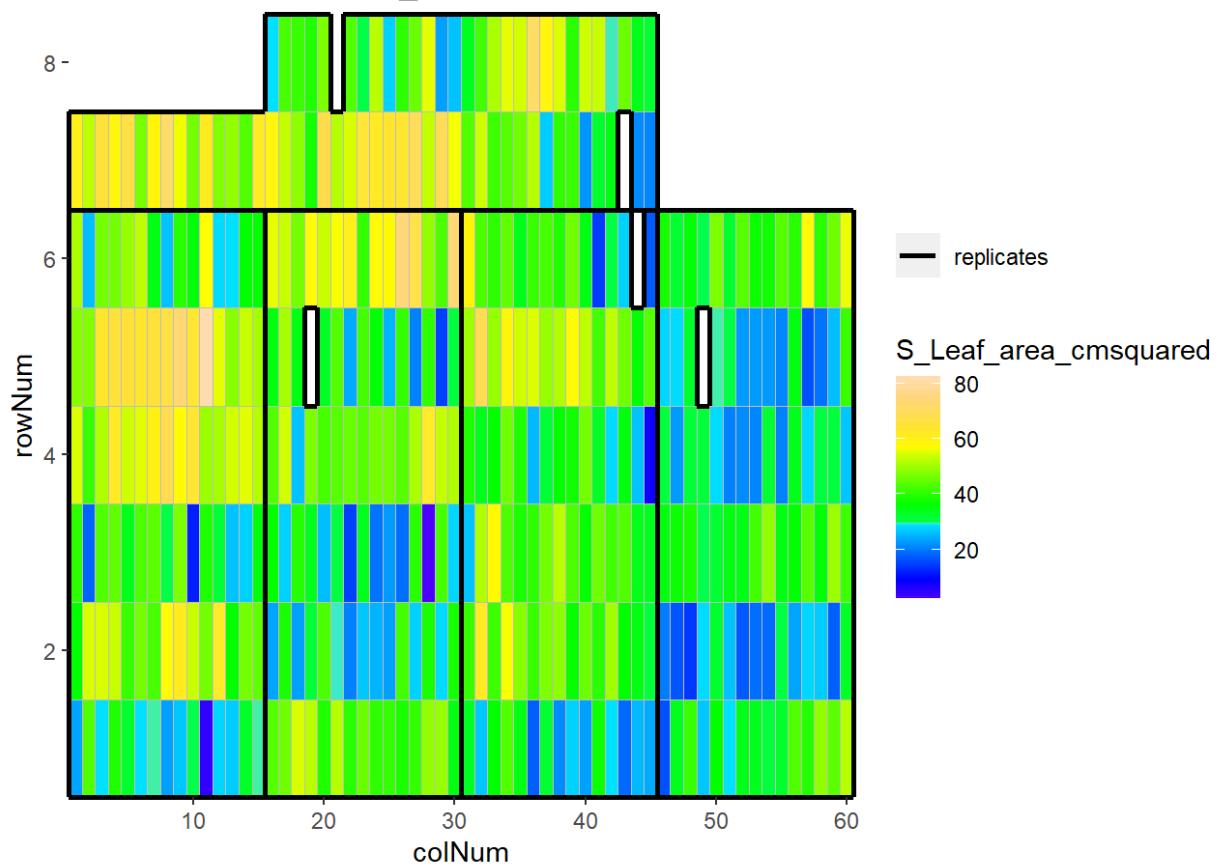
## EPPN2020\_M3P - 2020-03-01



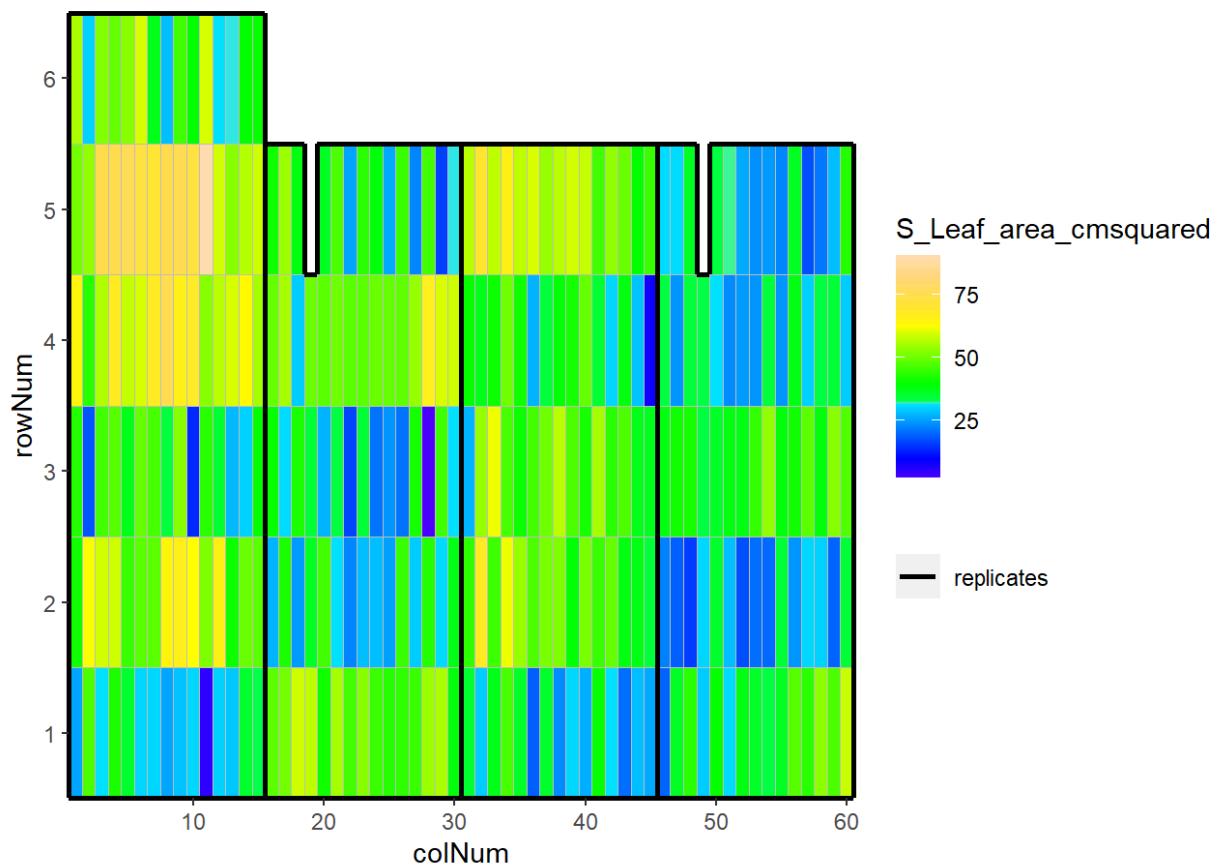
## EPPN2020\_M3P - 2020-03-02



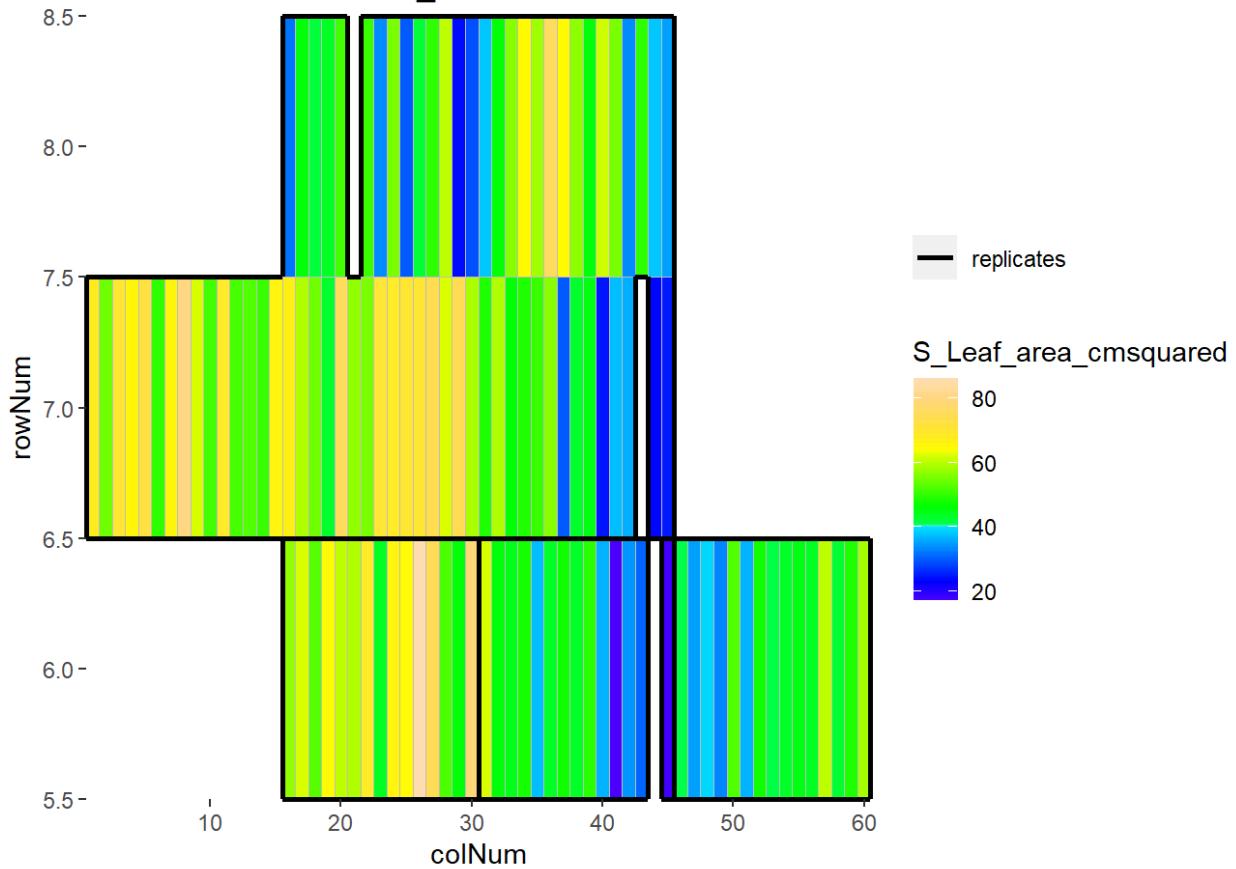
## EPPN2020\_M3P - 2020-03-03



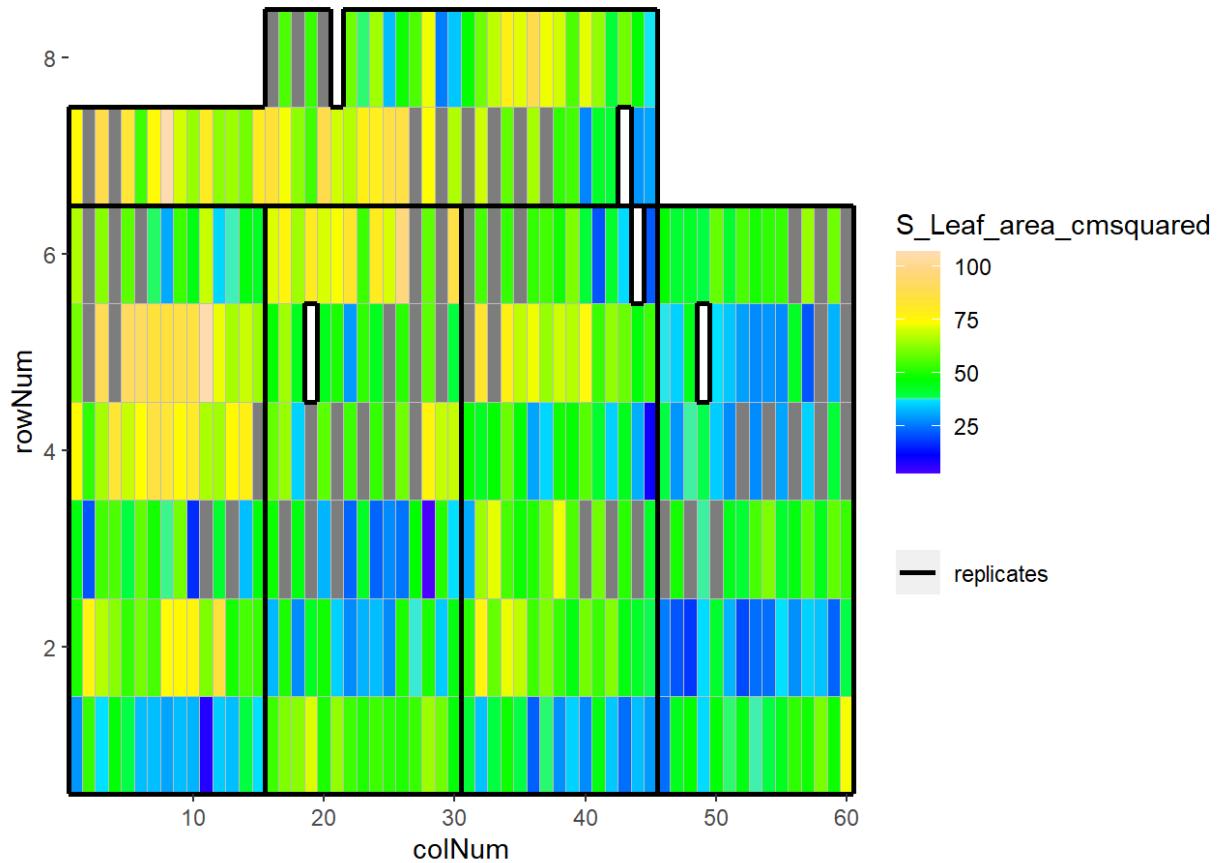
## EPPN2020\_M3P - 2020-03-04



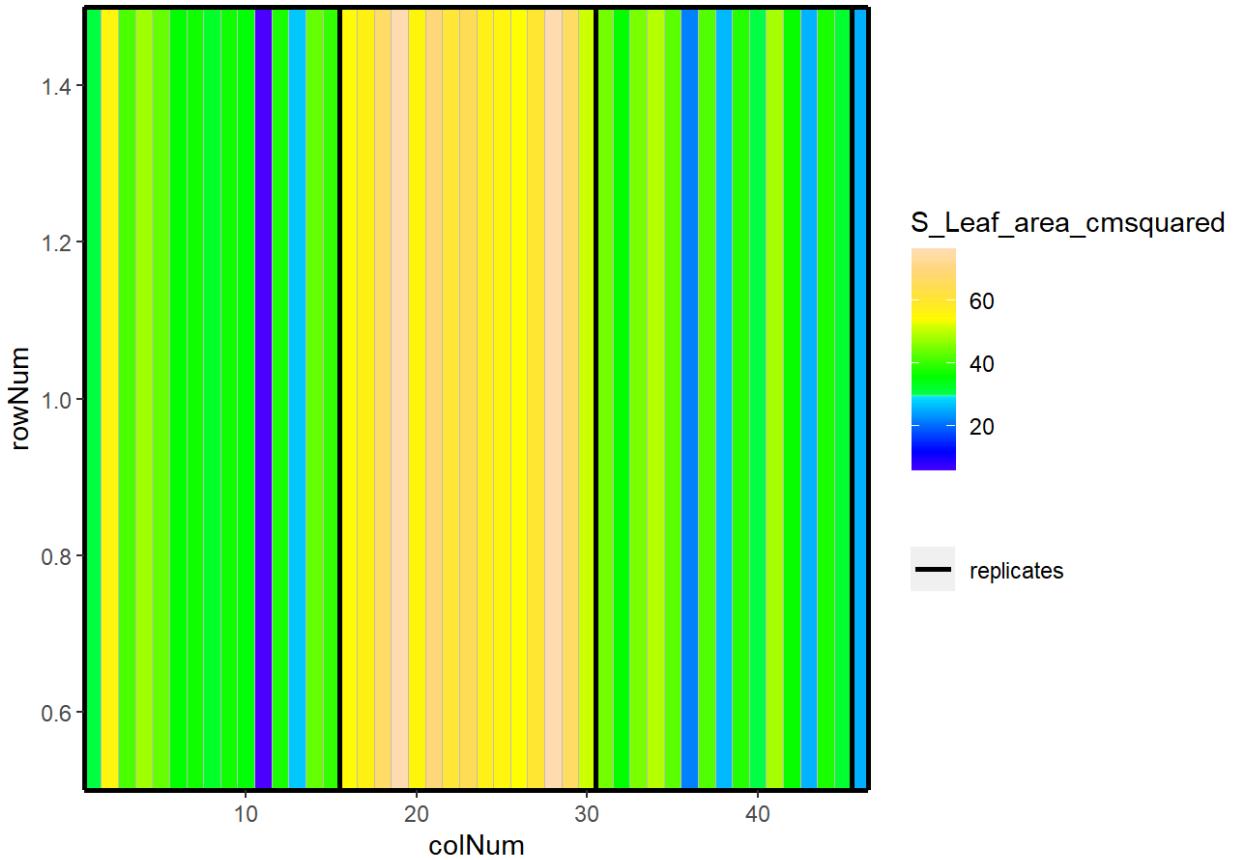
## EPPN2020\_M3P - 2020-03-05



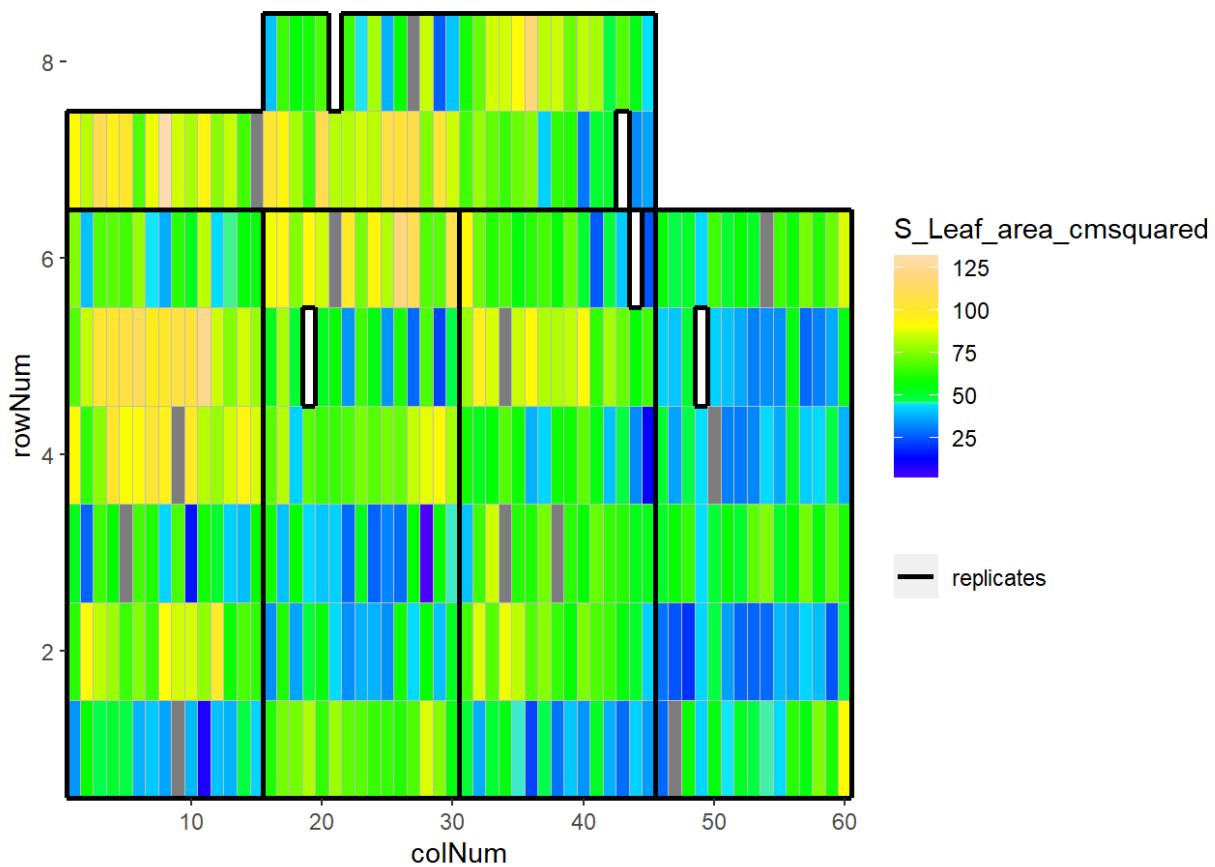
## EPPN2020\_M3P - 2020-03-06



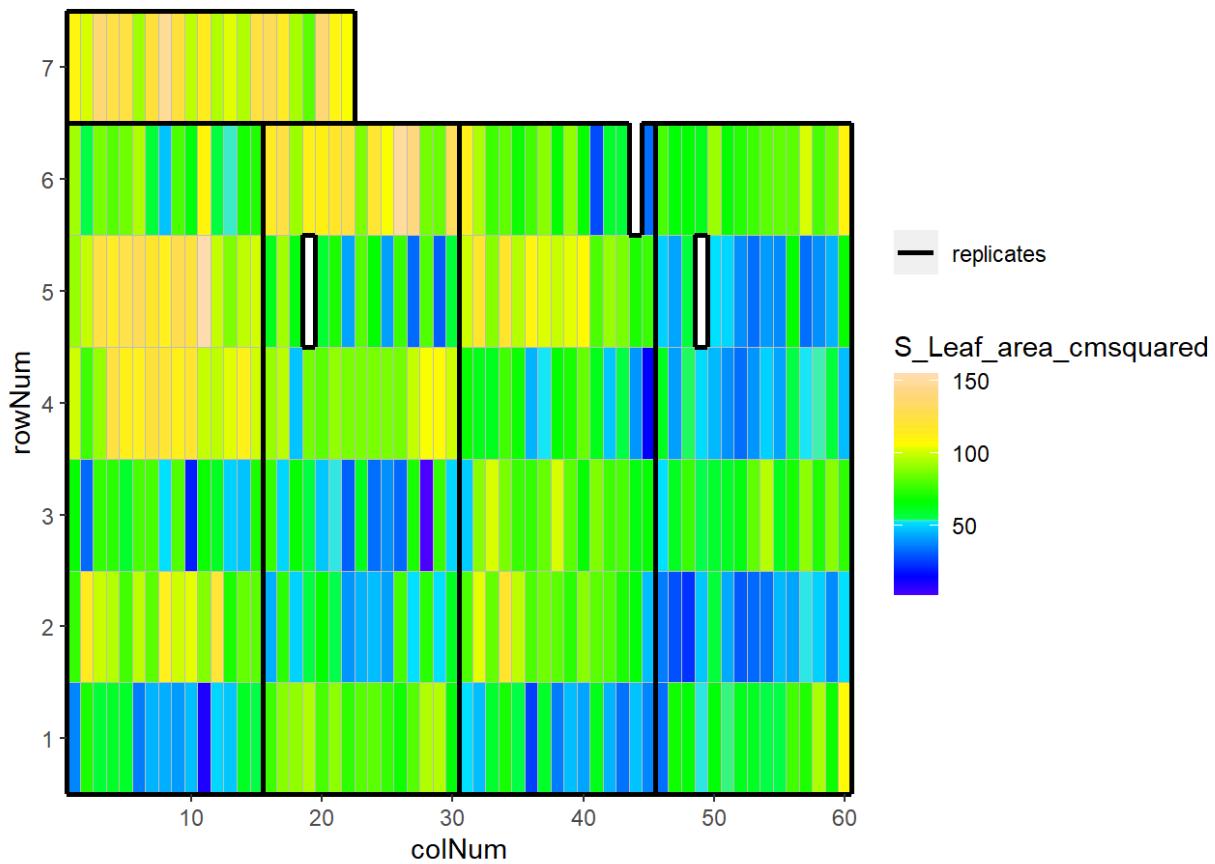
## EPPN2020\_M3P - 2020-03-07



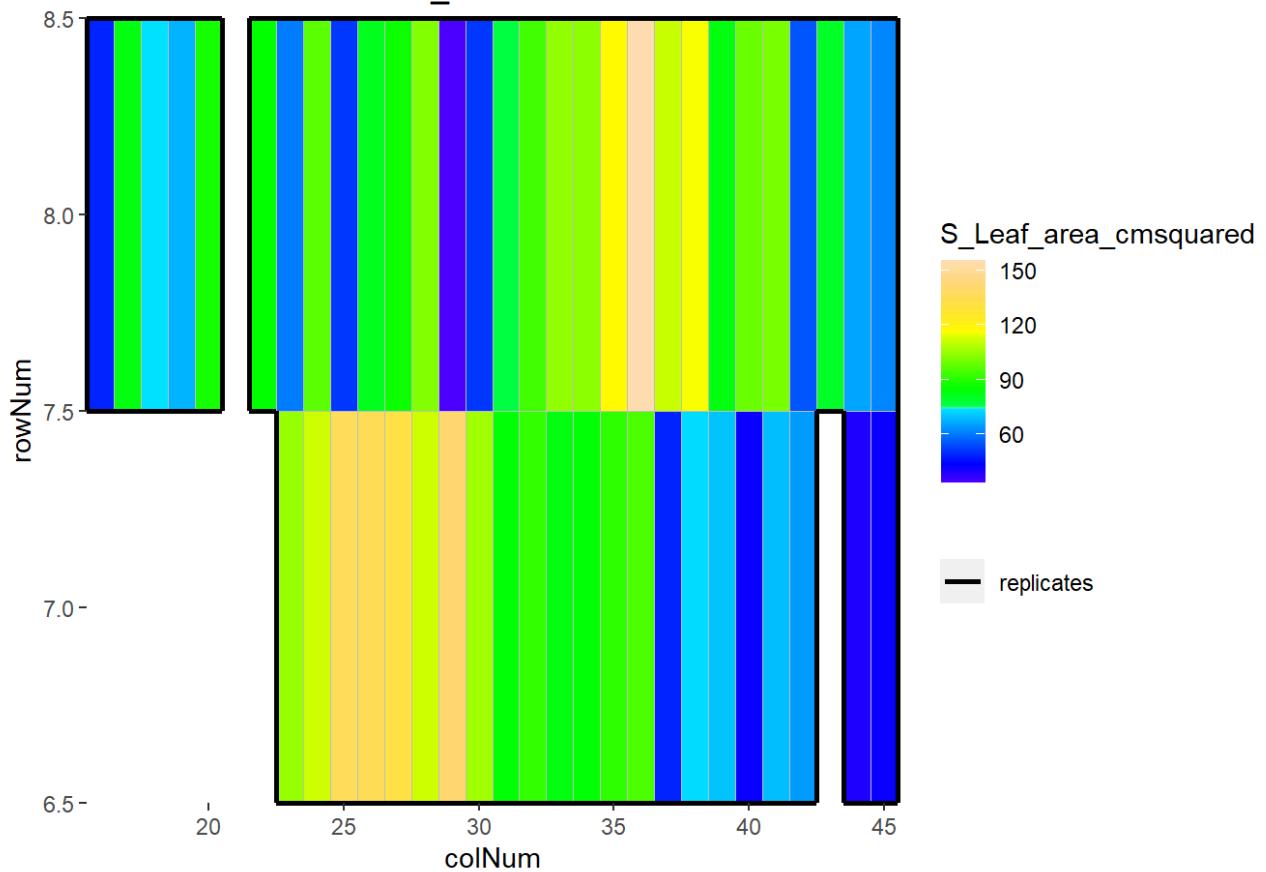
## EPPN2020\_M3P - 2020-03-08



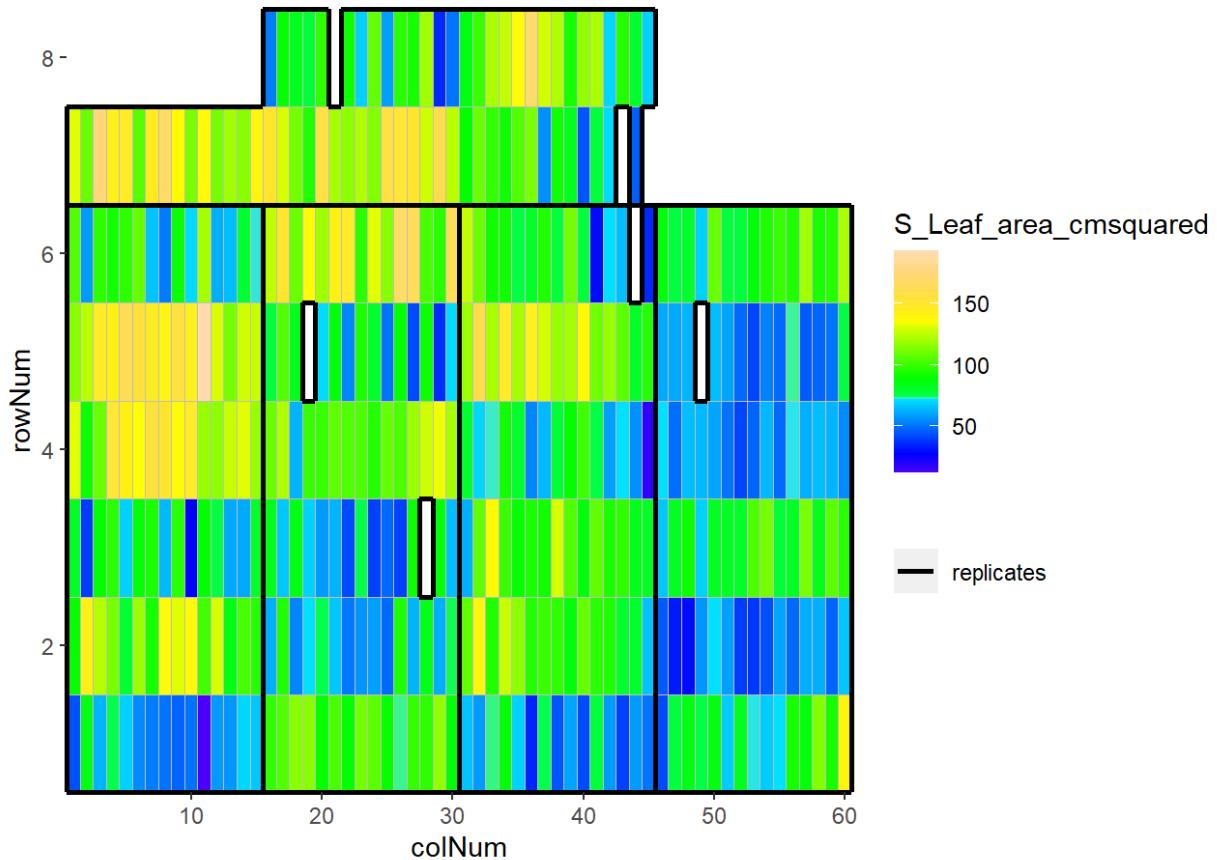
## EPPN2020\_M3P - 2020-03-11



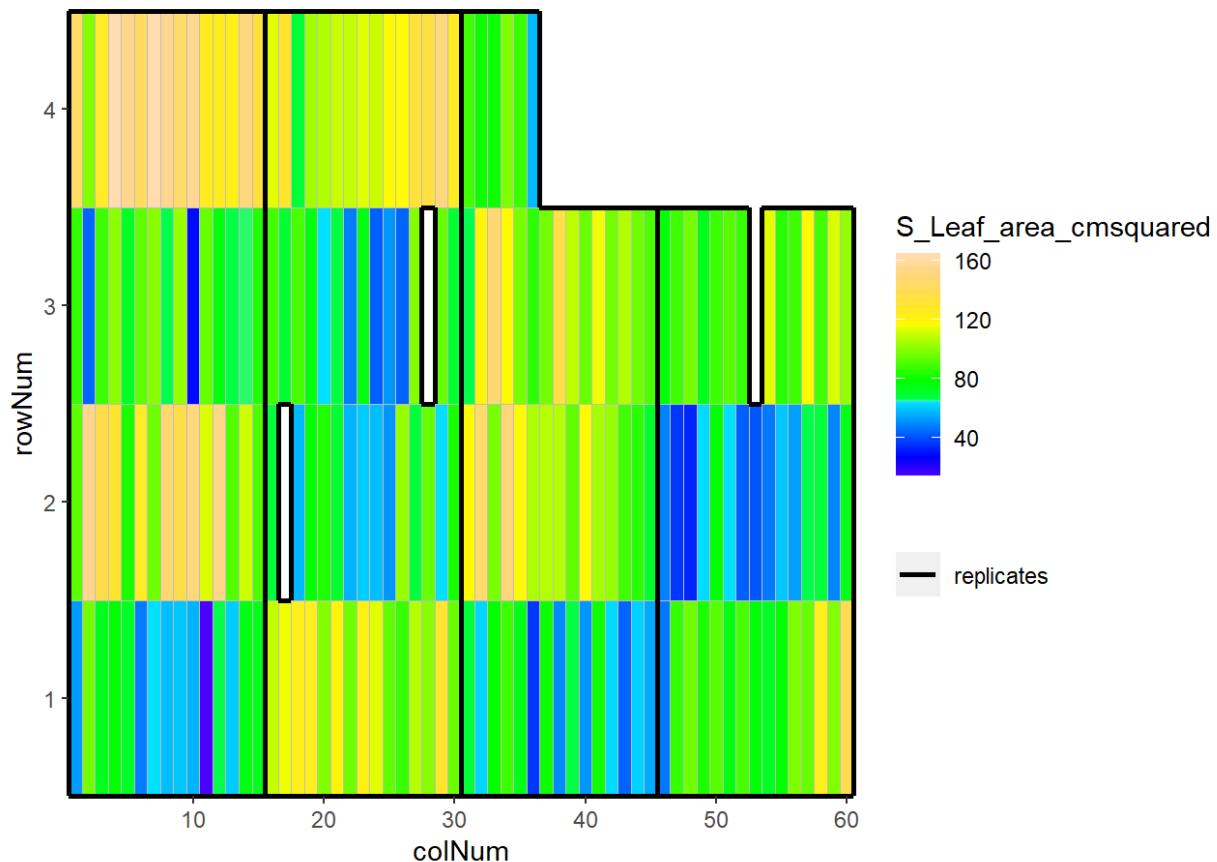
## EPPN2020\_M3P - 2020-03-12



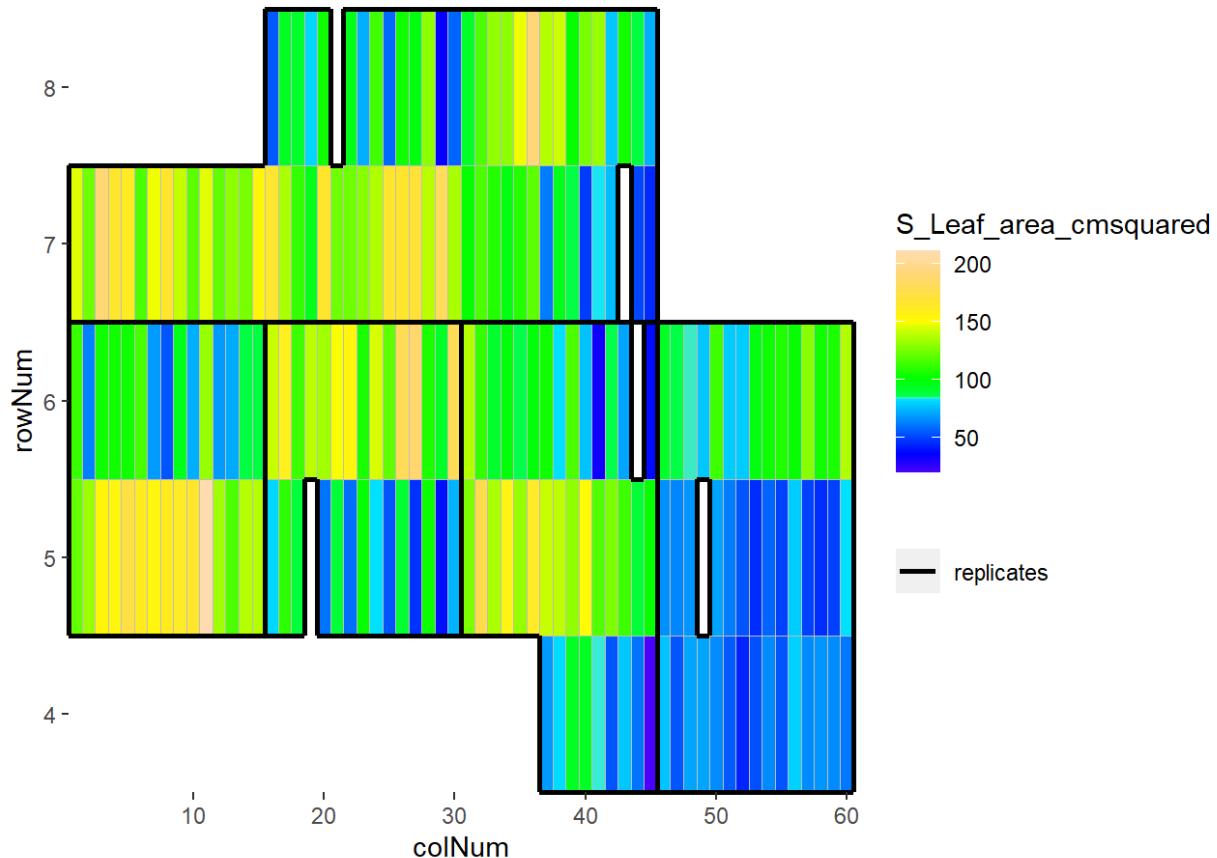
## EPPN2020\_M3P - 2020-03-13



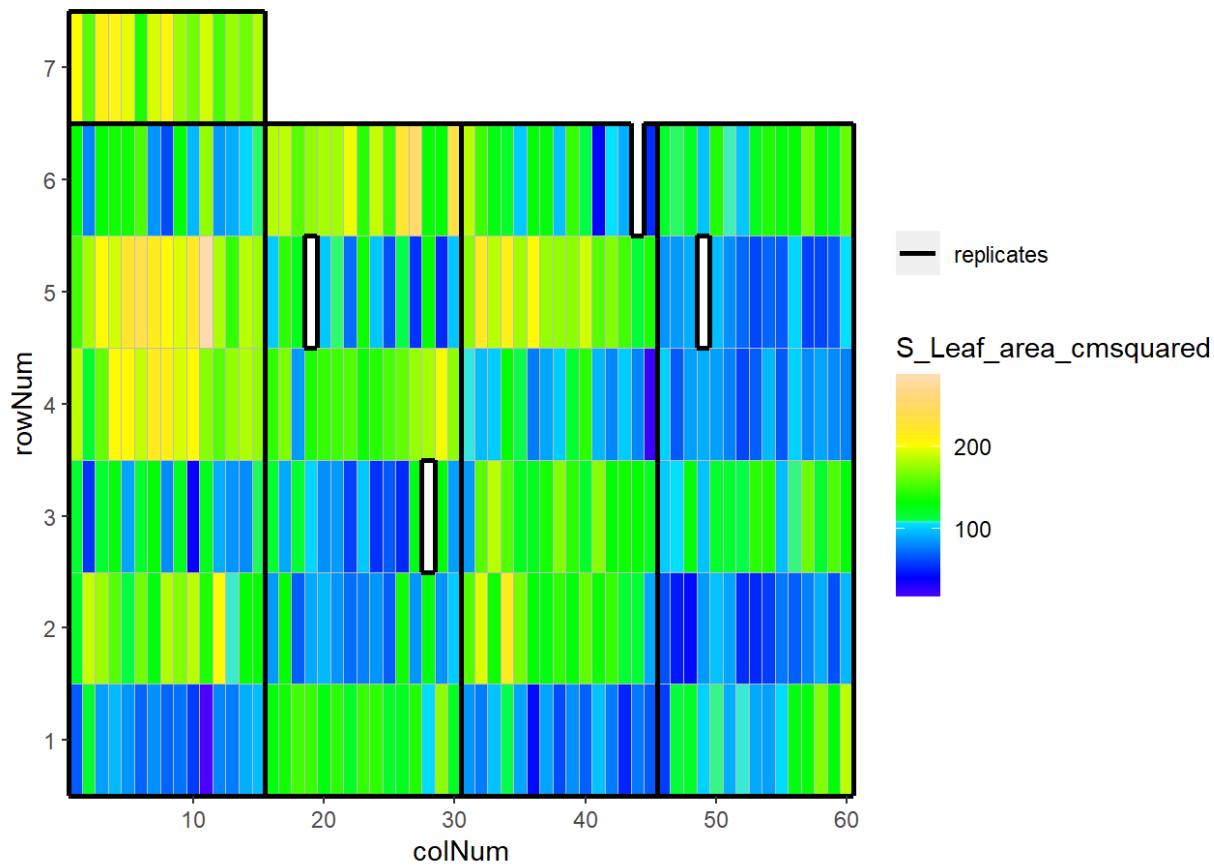
## EPPN2020\_M3P - 2020-03-14



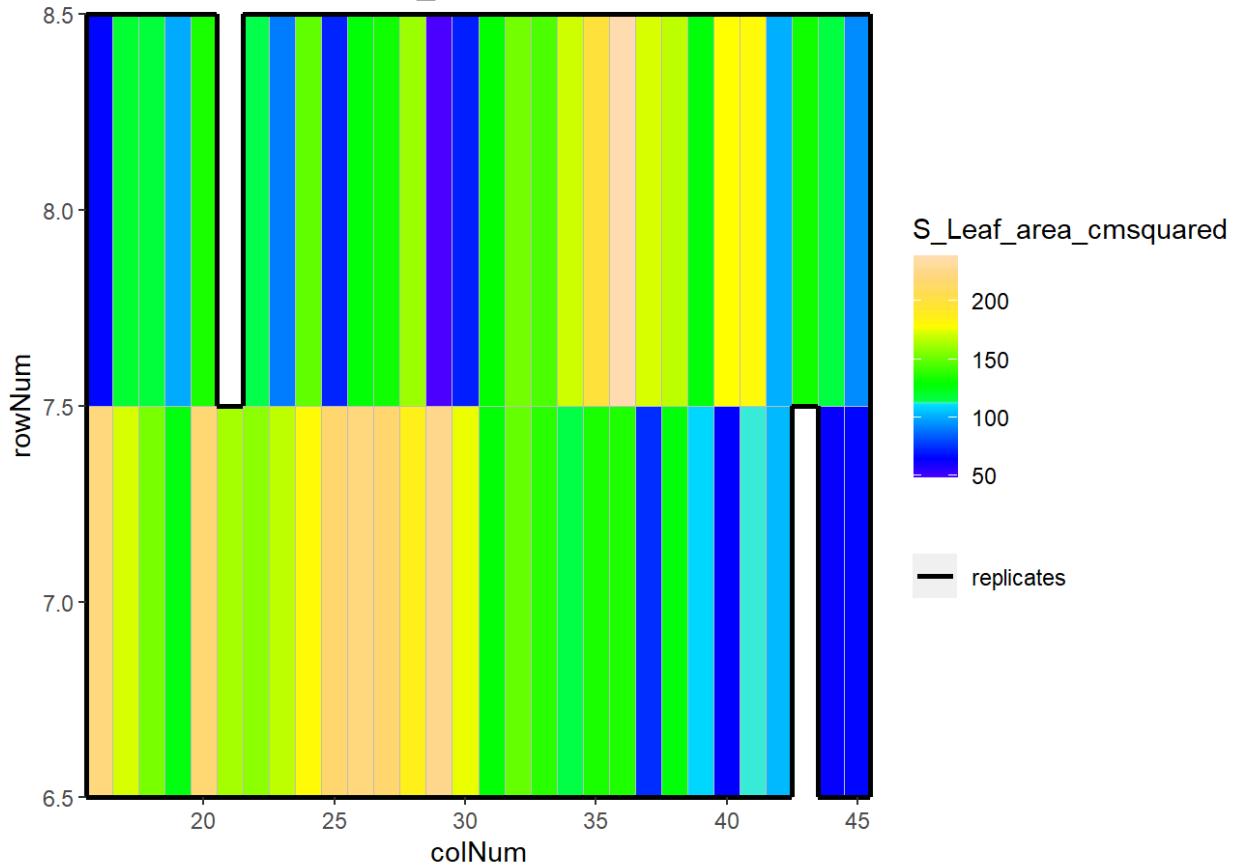
## EPPN2020\_M3P - 2020-03-15



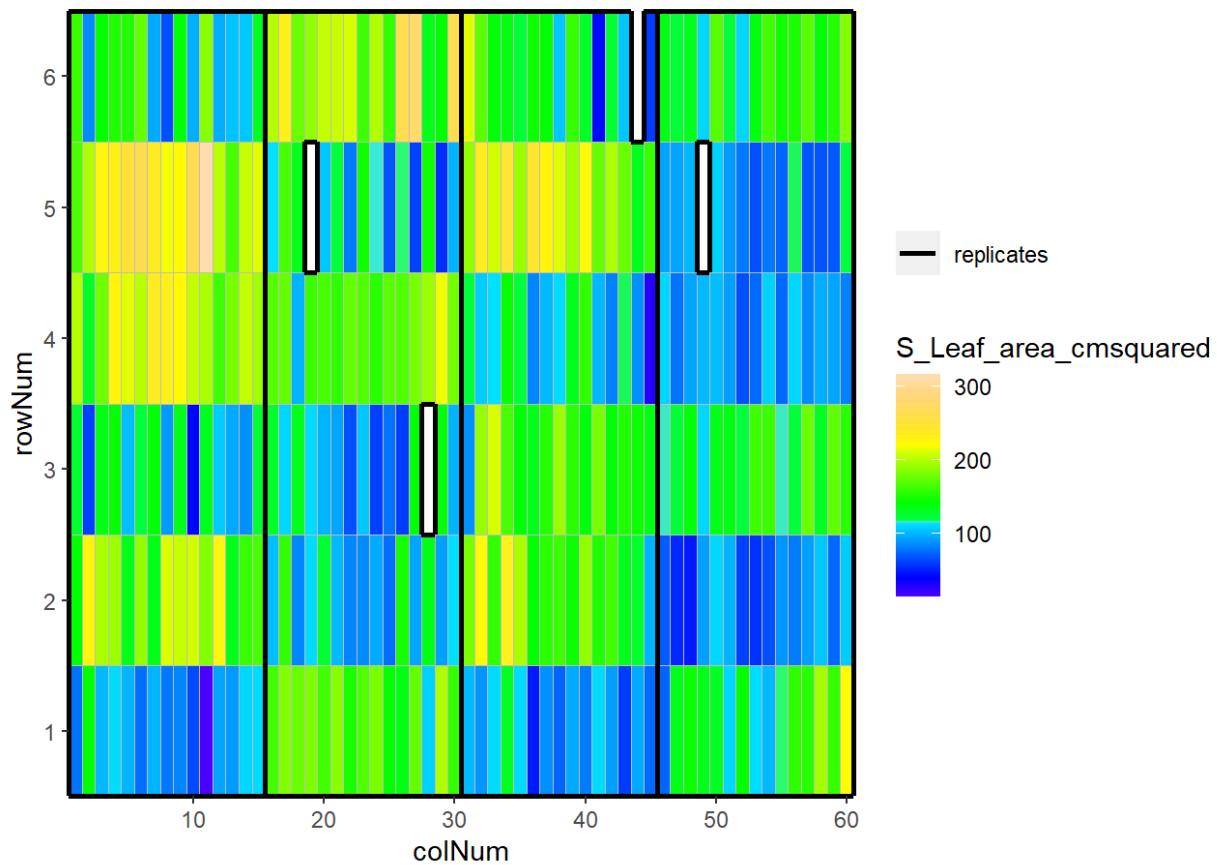
## EPPN2020\_M3P - 2020-03-18



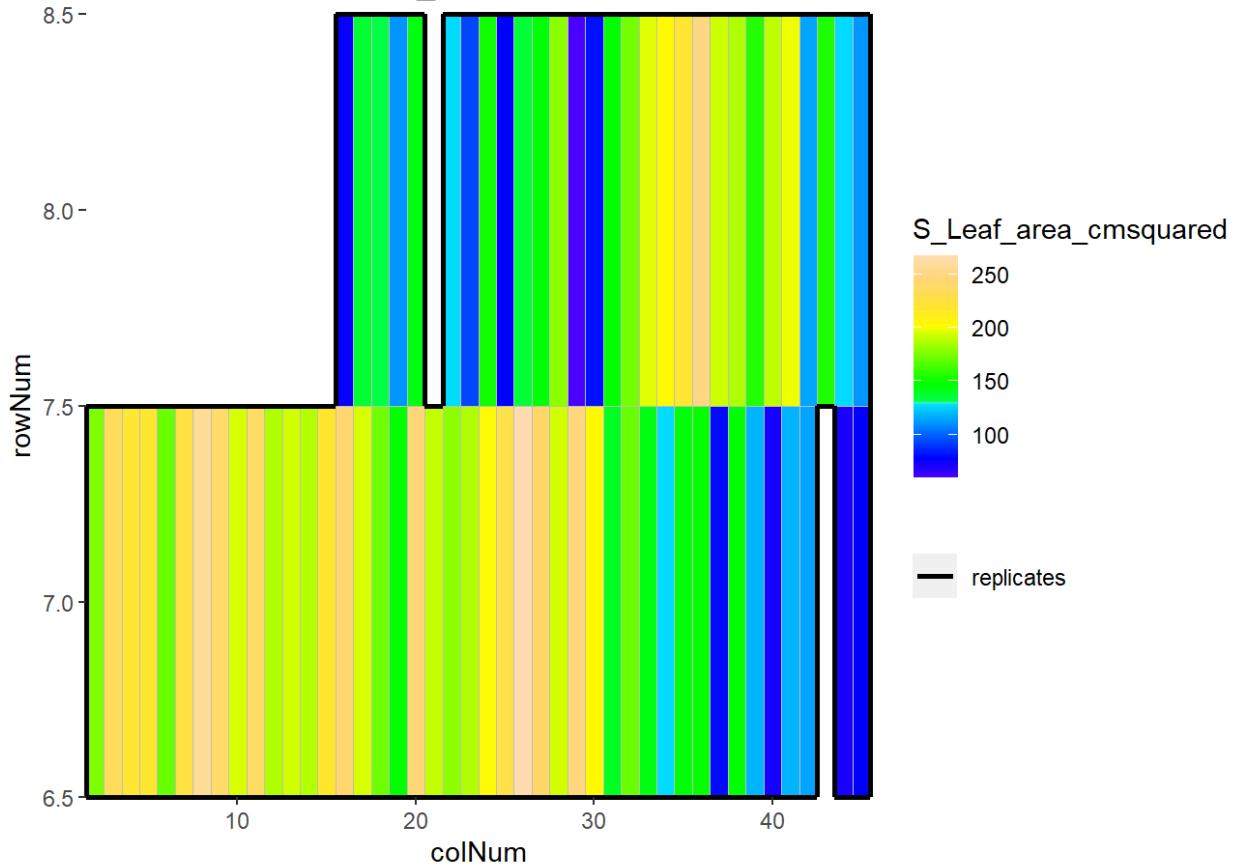
## EPPN2020\_M3P - 2020-03-19



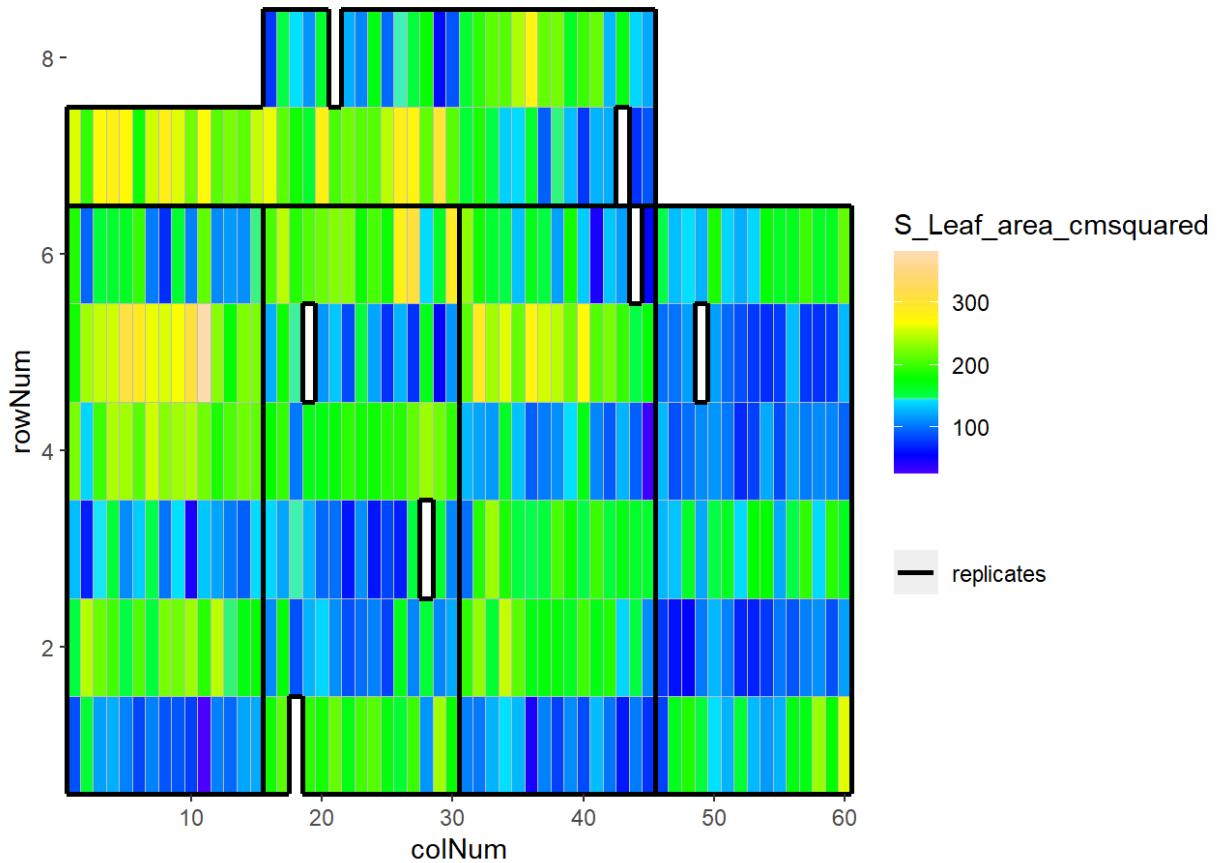
## EPPN2020\_M3P - 2020-03-20



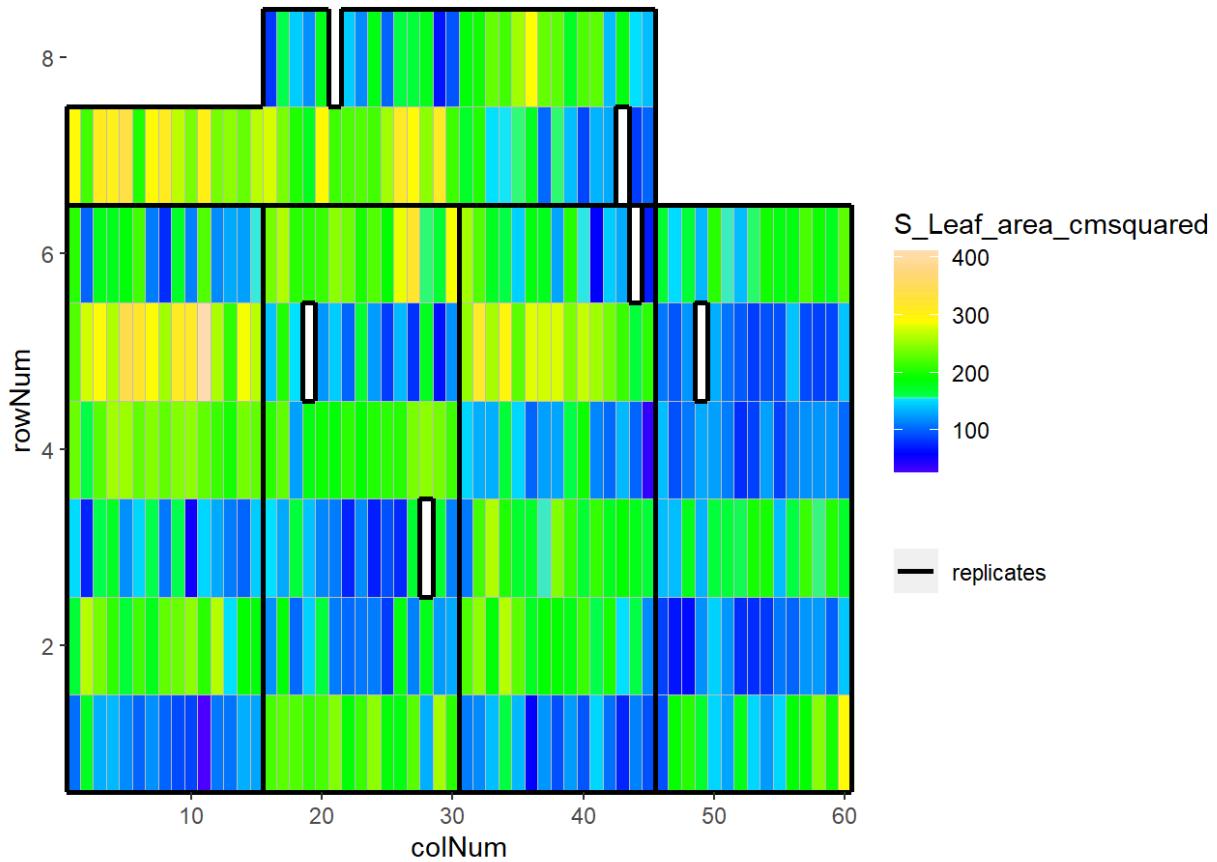
## EPPN2020\_M3P - 2020-03-21



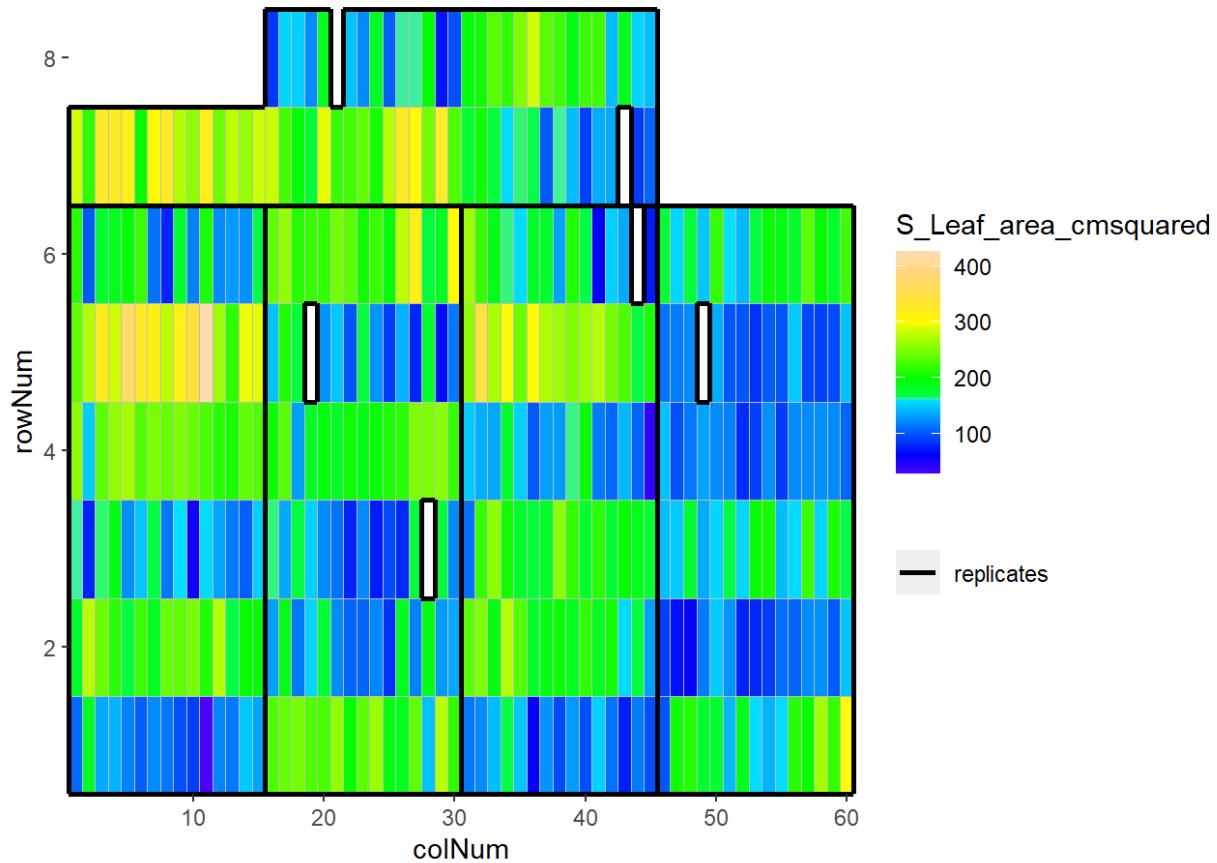
## EPPN2020\_M3P - 2020-03-22



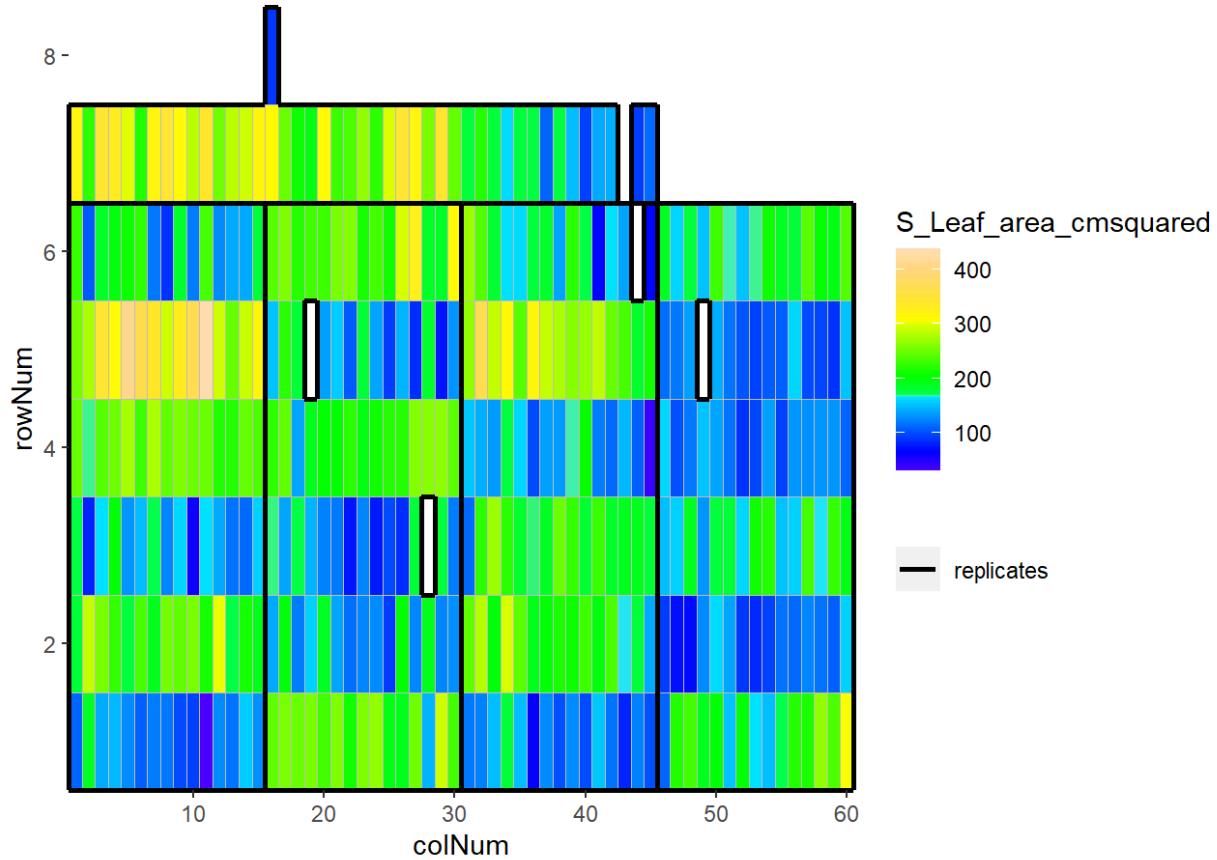
## EPPN2020\_M3P - 2020-03-24



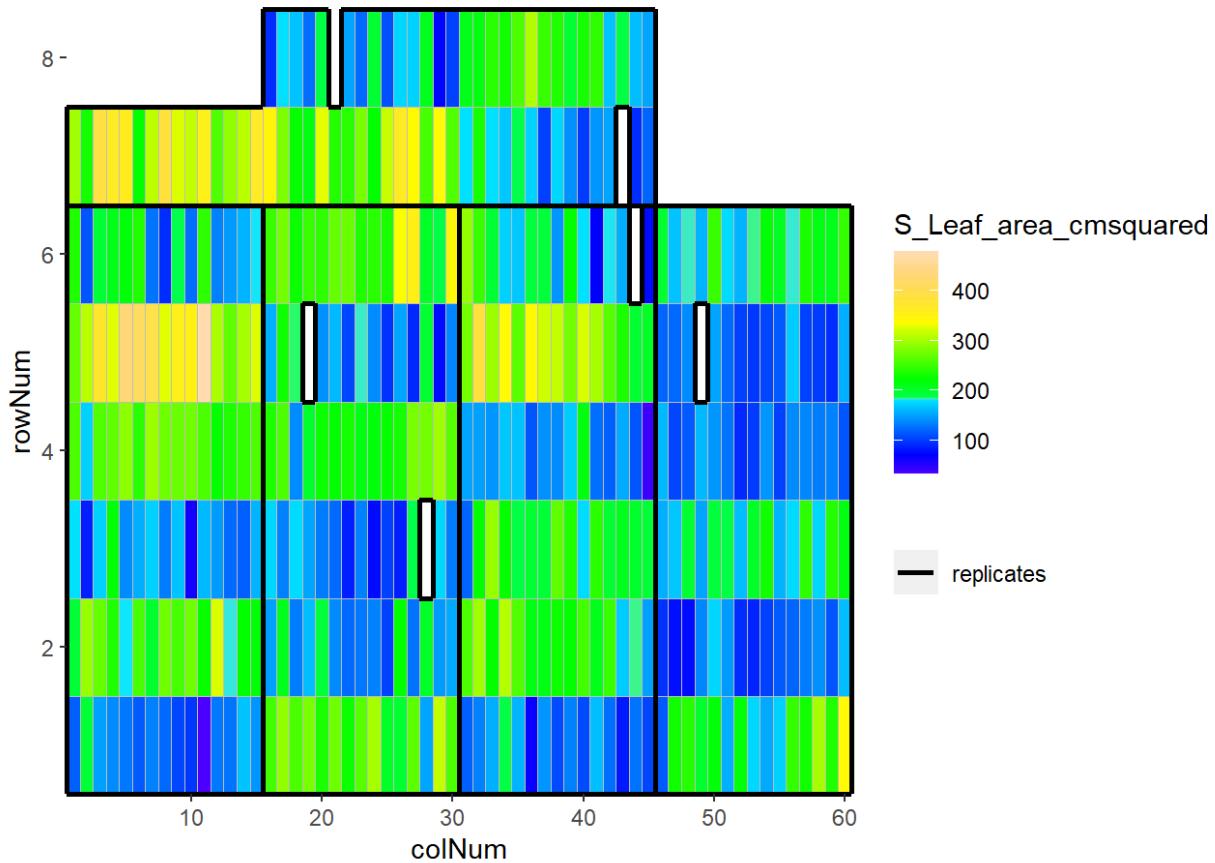
## EPPN2020\_M3P - 2020-03-25



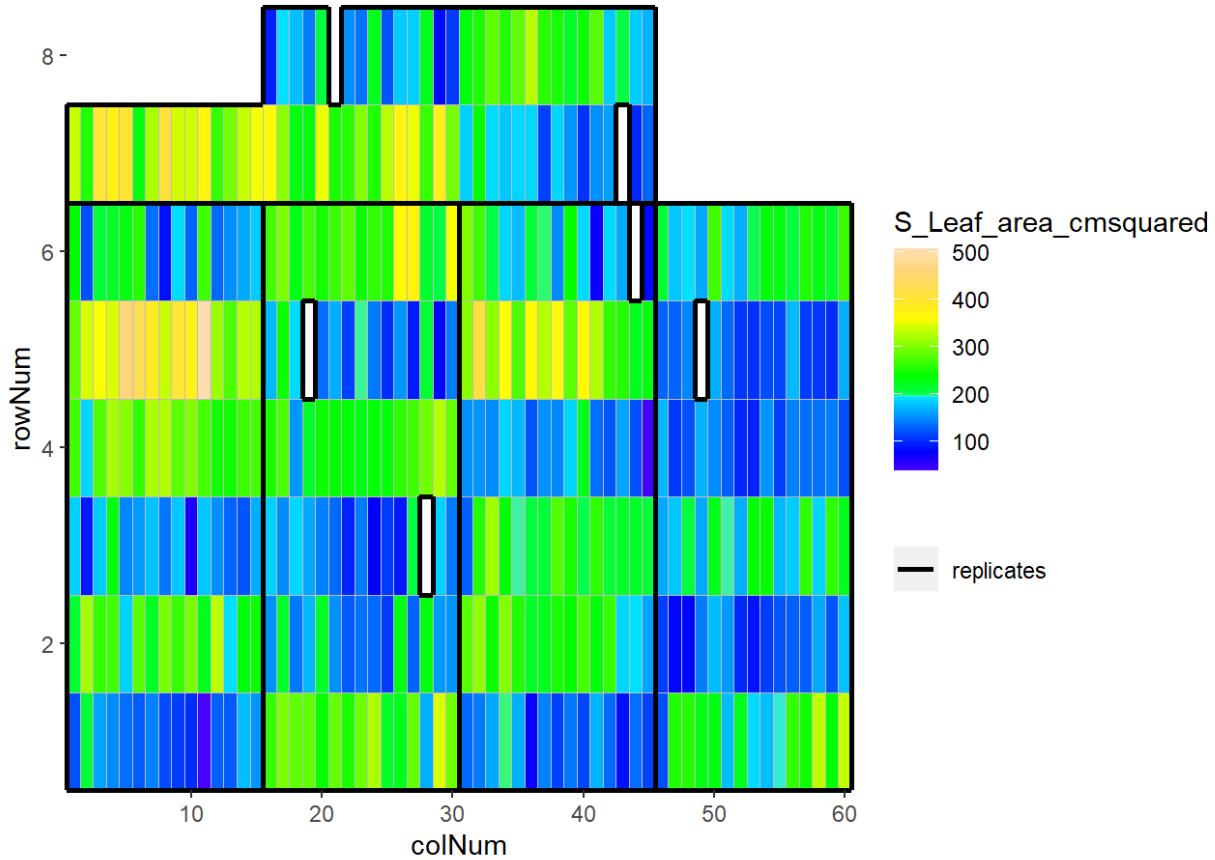
## EPPN2020\_M3P - 2020-03-26



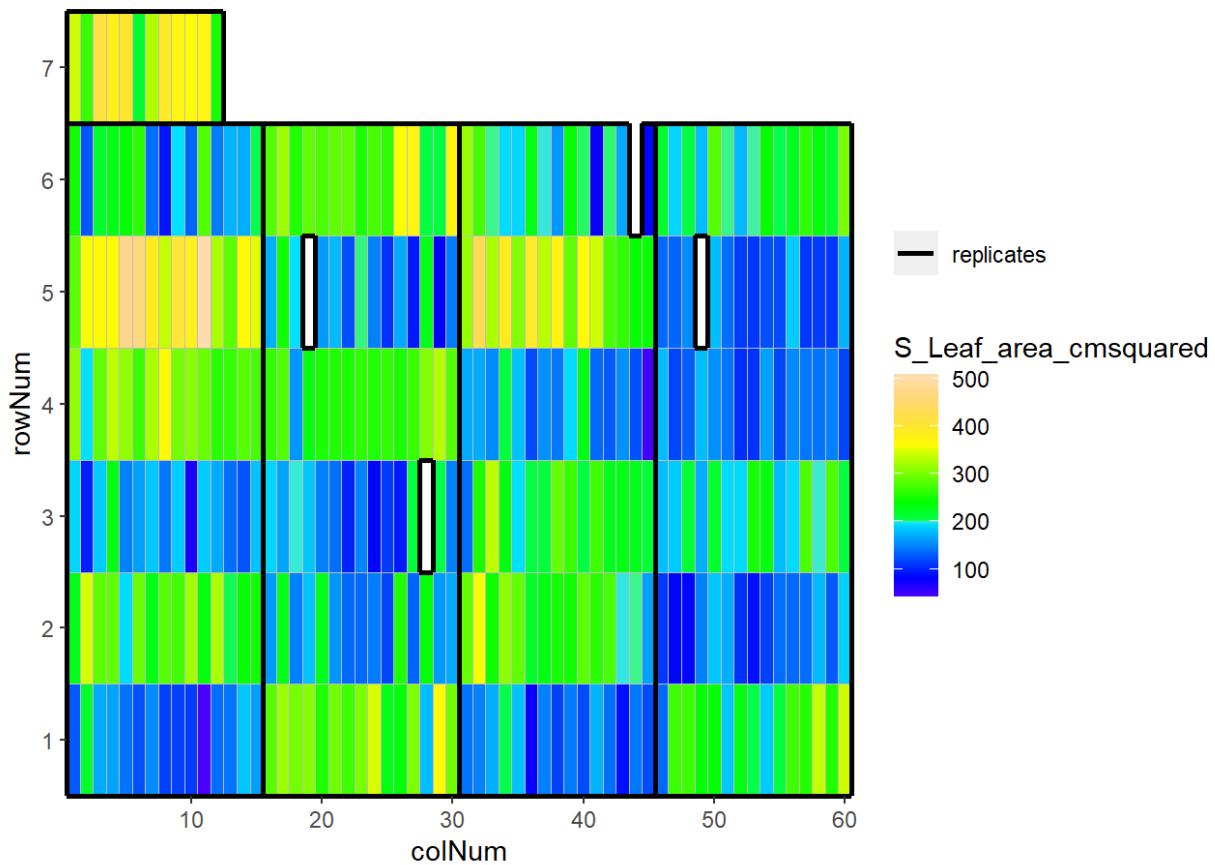
## EPPN2020\_M3P - 2020-03-27



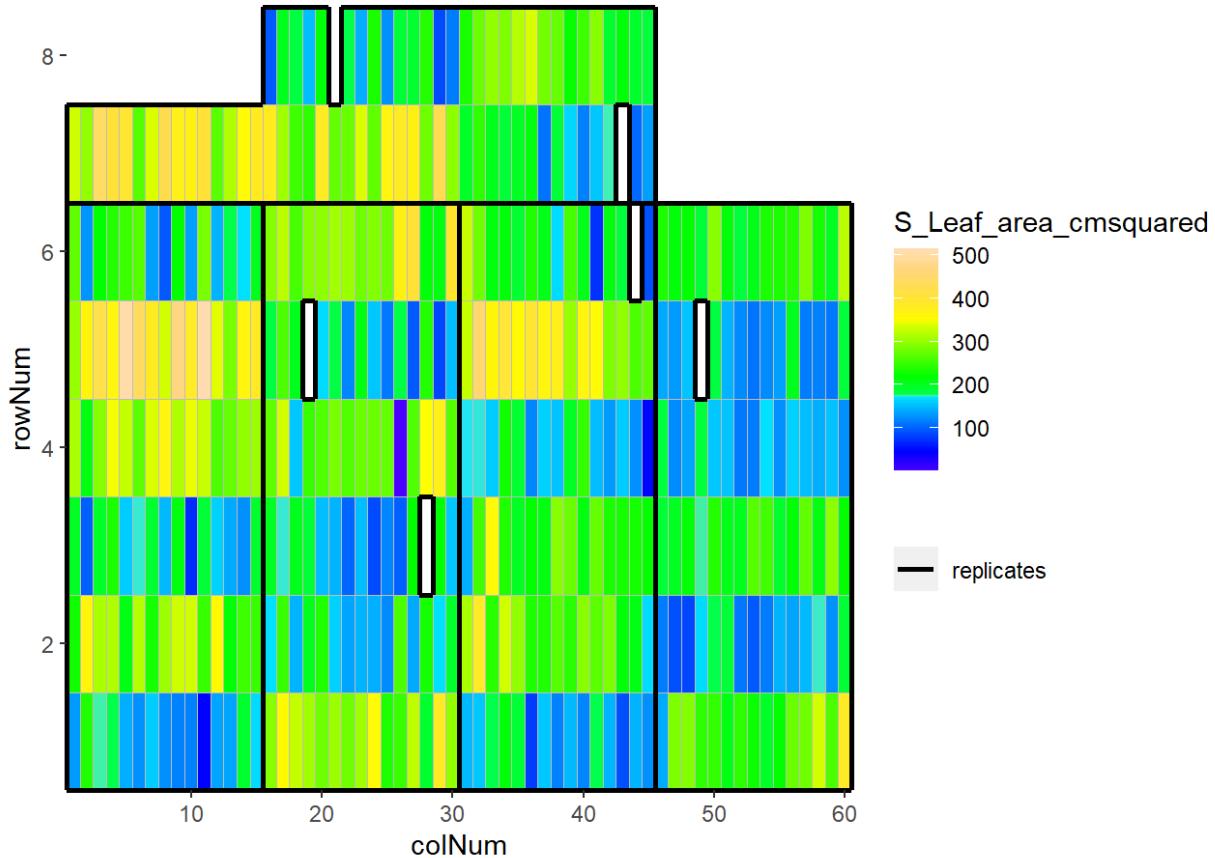
## EPPN2020\_M3P - 2020-03-28



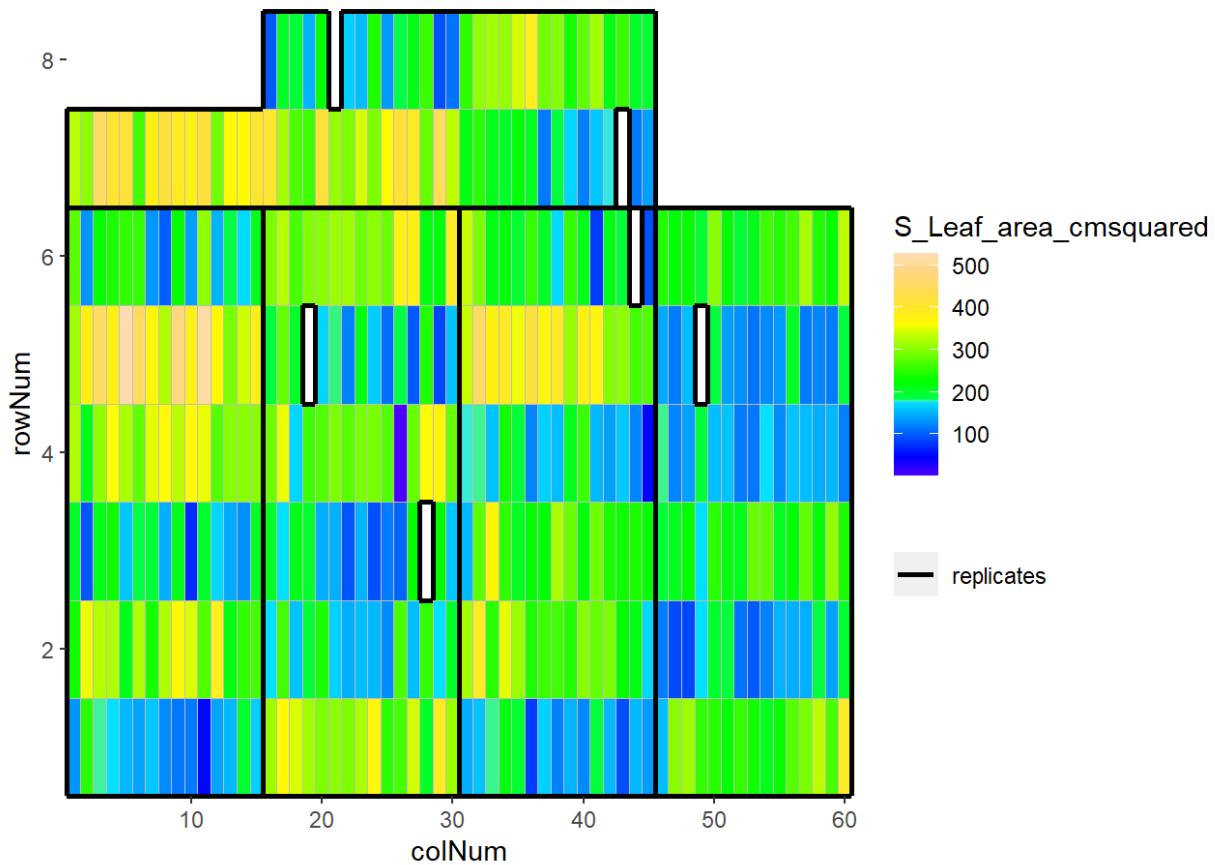
## EPPN2020\_M3P - 2020-03-29



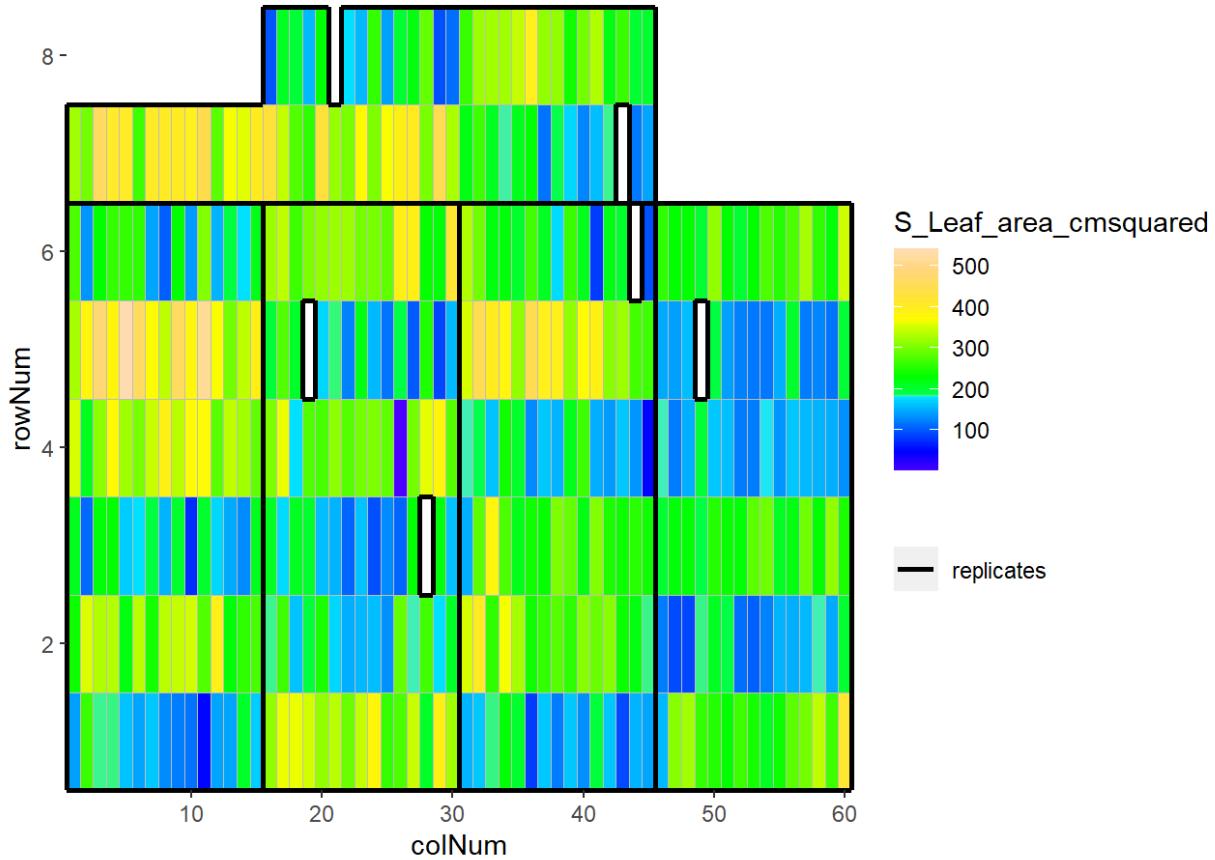
## EPPN2020\_M3P - 2020-03-30



## EPPN2020\_M3P - 2020-03-31



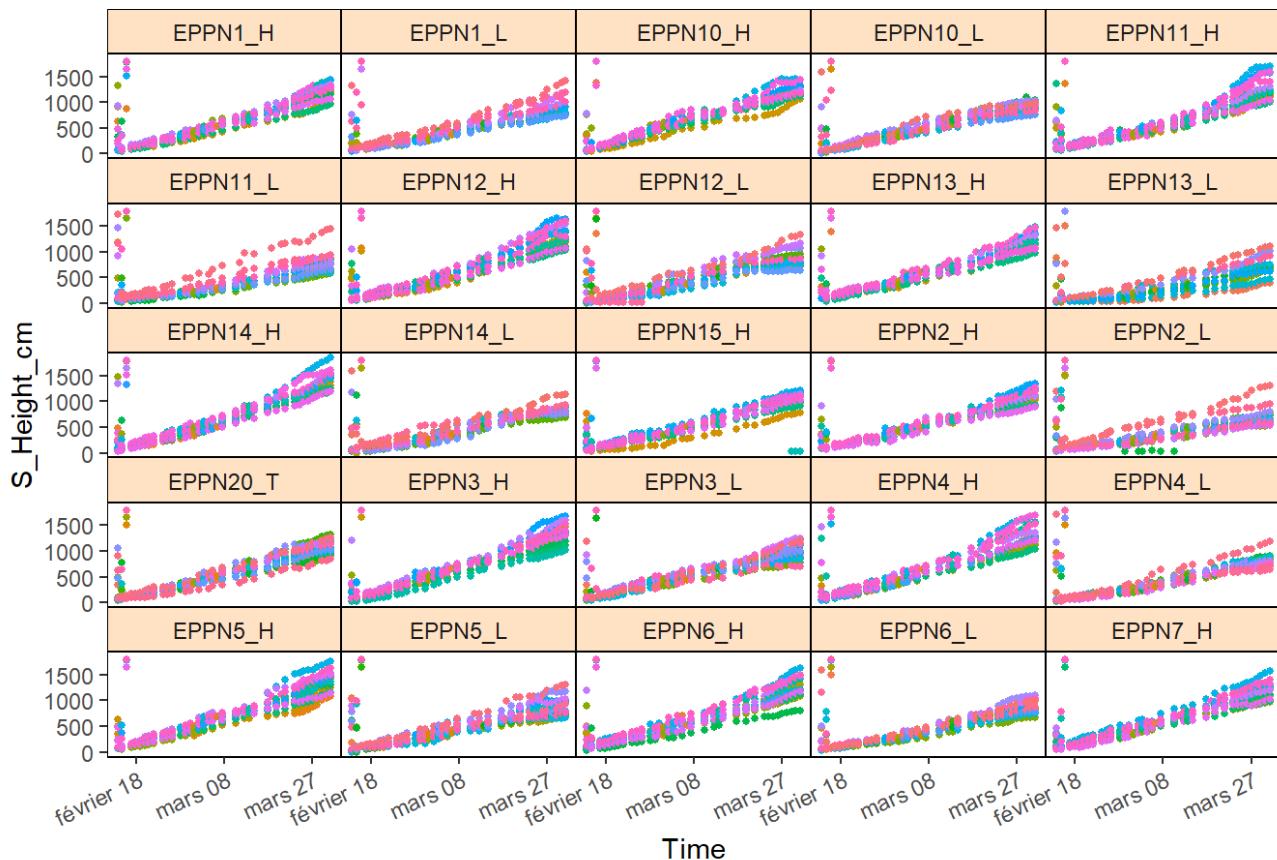
## EPPN2020\_M3P - 2020-04-01



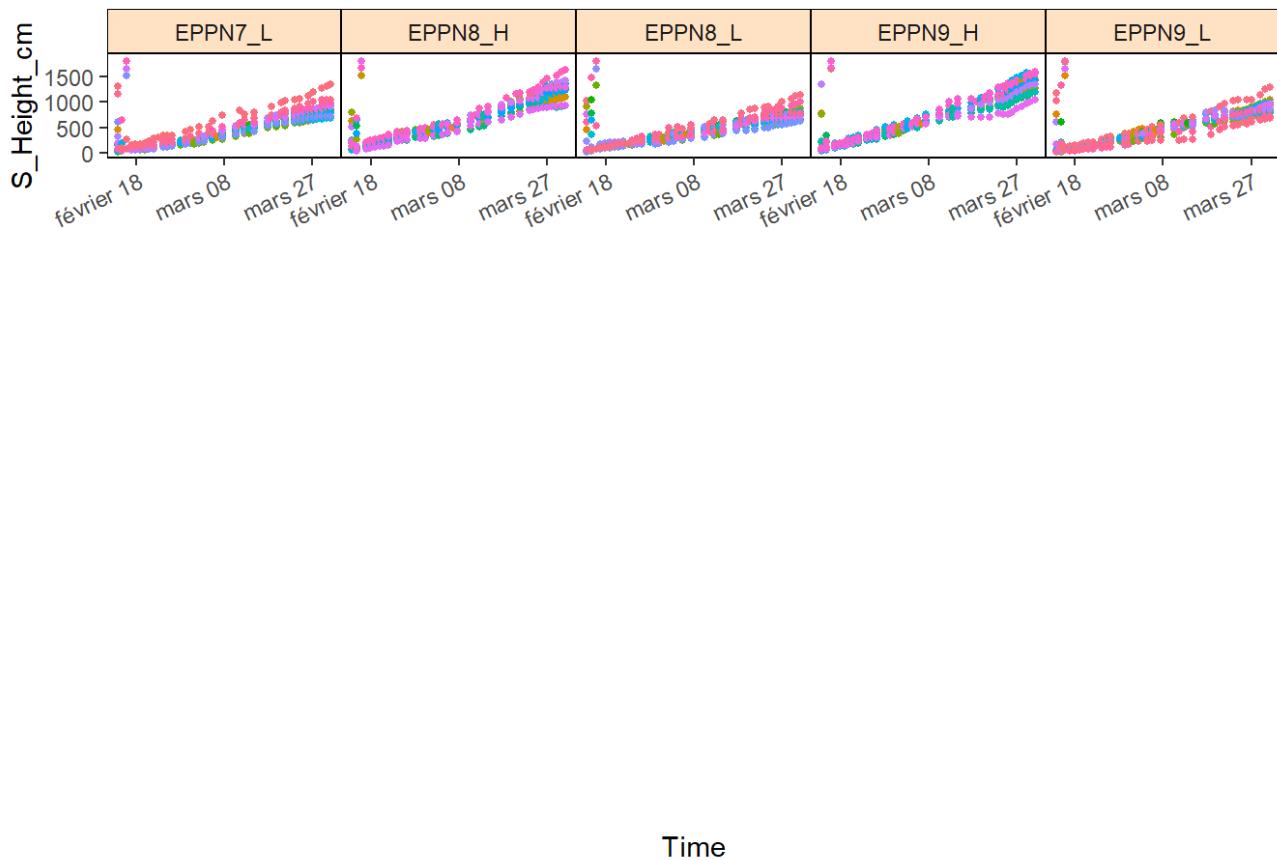
Time course, boxplots and correlation plots of data

```
for (trait_name in traits) {  
  single_outliers_name <- paste0("Single_outliers_", trait_name)  
  
  if (exists(single_outliers_name)) {  
    Single_outliers <- get(single_outliers_name)  
  
    plot(Single_outliers,  
          traits = trait_name,  
          plotType = "raw")  
  
    plot(Single_outliers,  
          plotType = "box",  
          traits = trait_name)  
  
    plot(Single_outliers,  
          plotType = "cor",  
          traits = trait_name)  
  
  } else {  
    cat("No Single_outliers object found for trait", trait_name, "\n")  
  }  
}
```

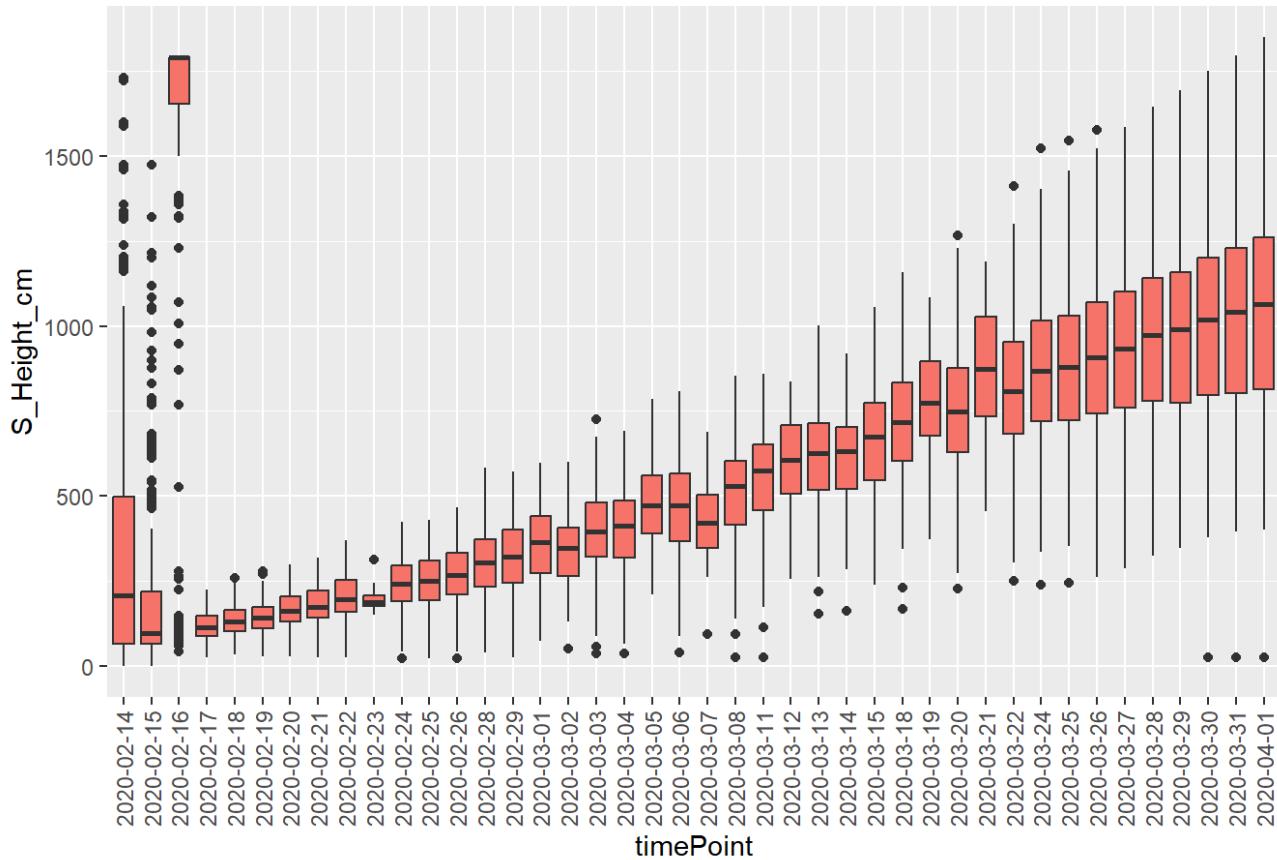
## EPPN2020\_M3P - S\_Height\_cm - raw data



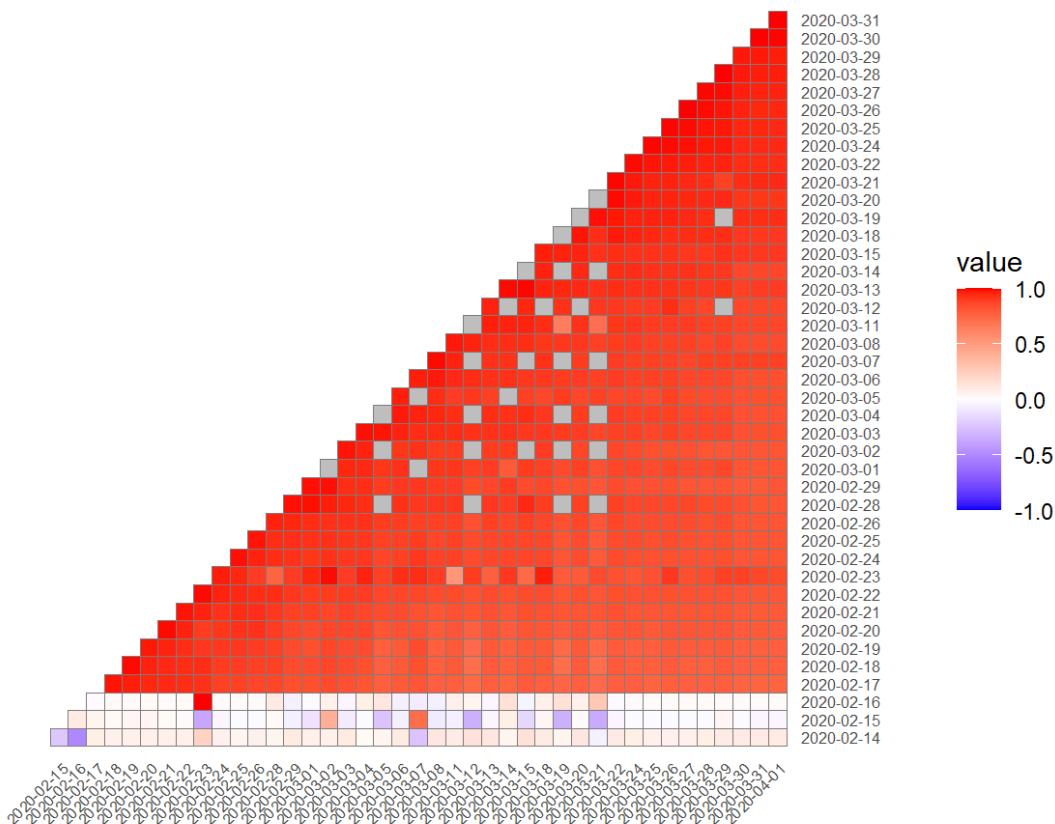
## EPPN2020\_M3P - S\_Height\_cm - raw data



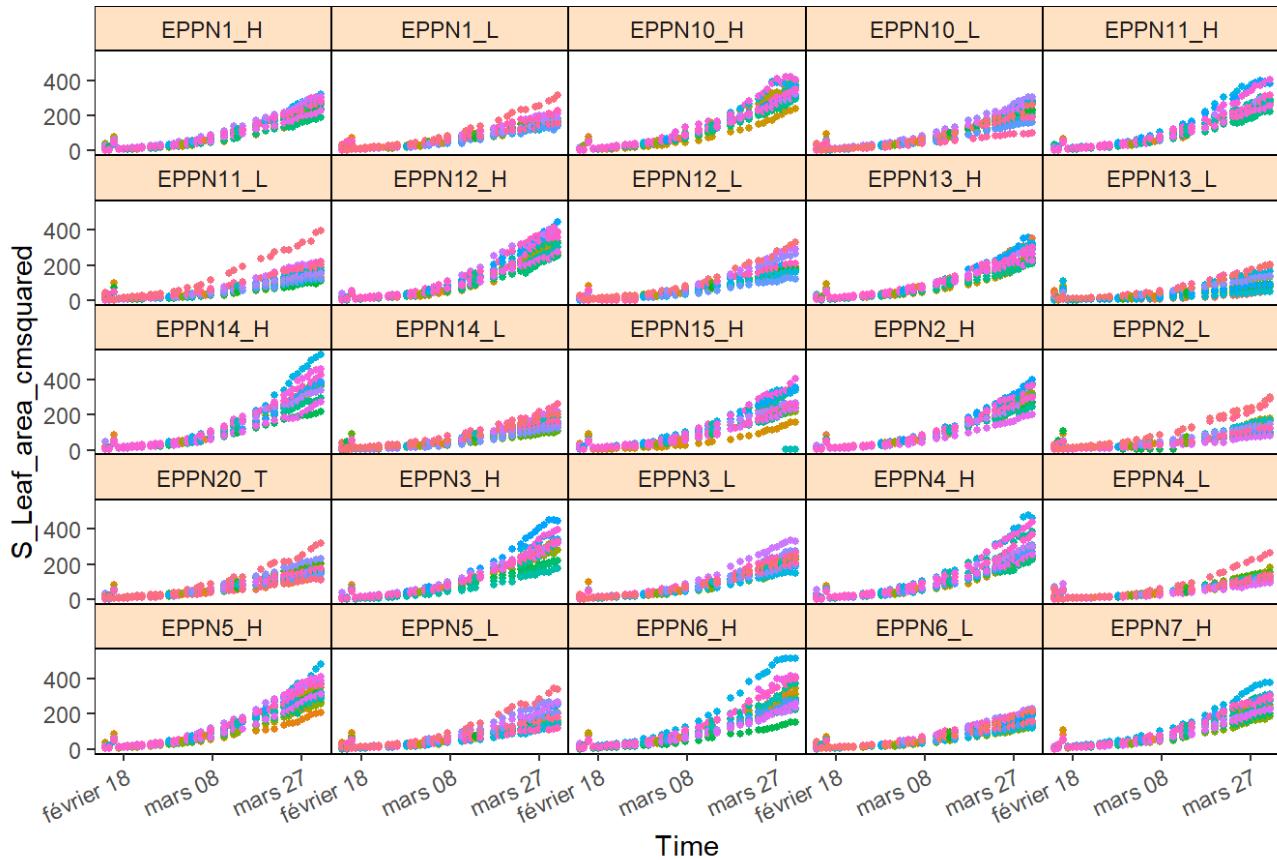
## EPPN2020\_M3P - S\_Height\_cm



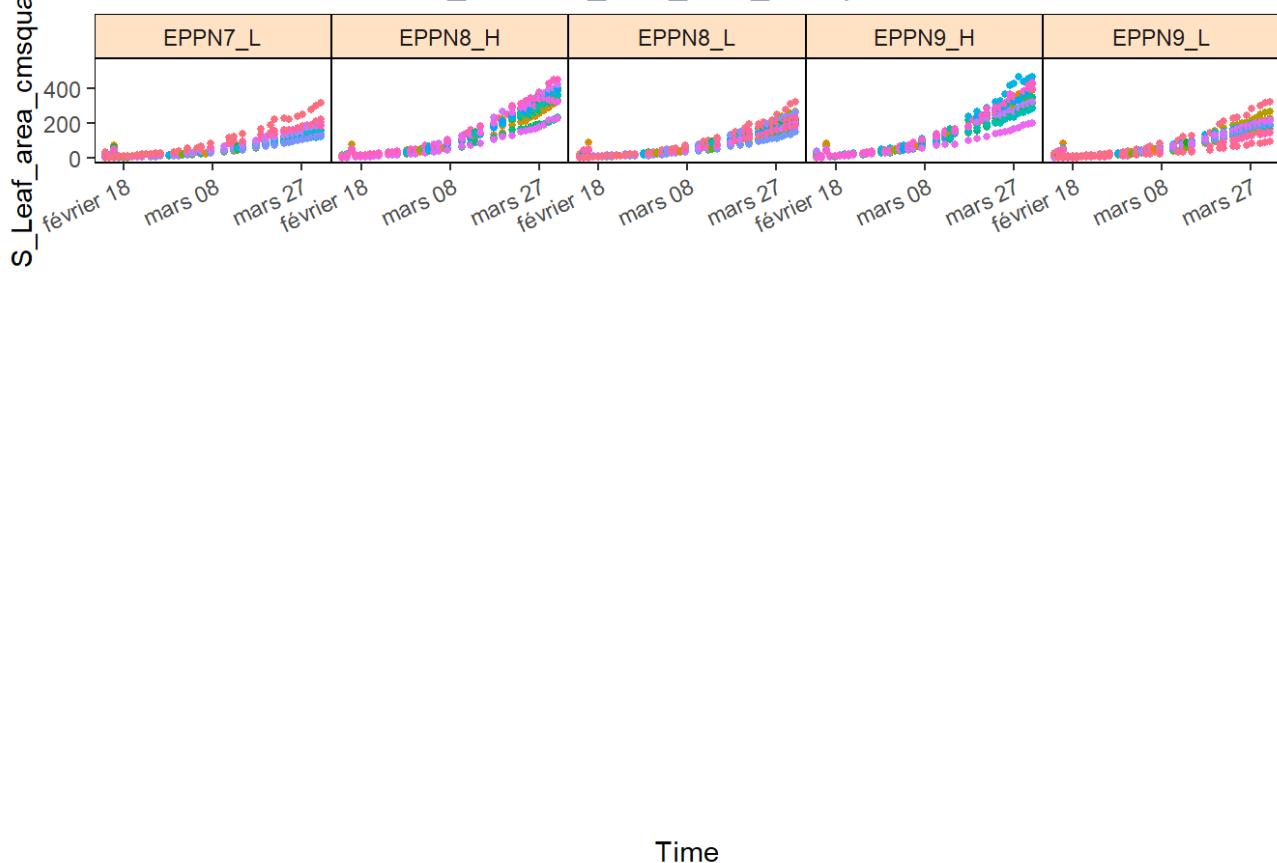
## EPPN2020\_M3P - Correlations of timepoints for S\_Height\_cm



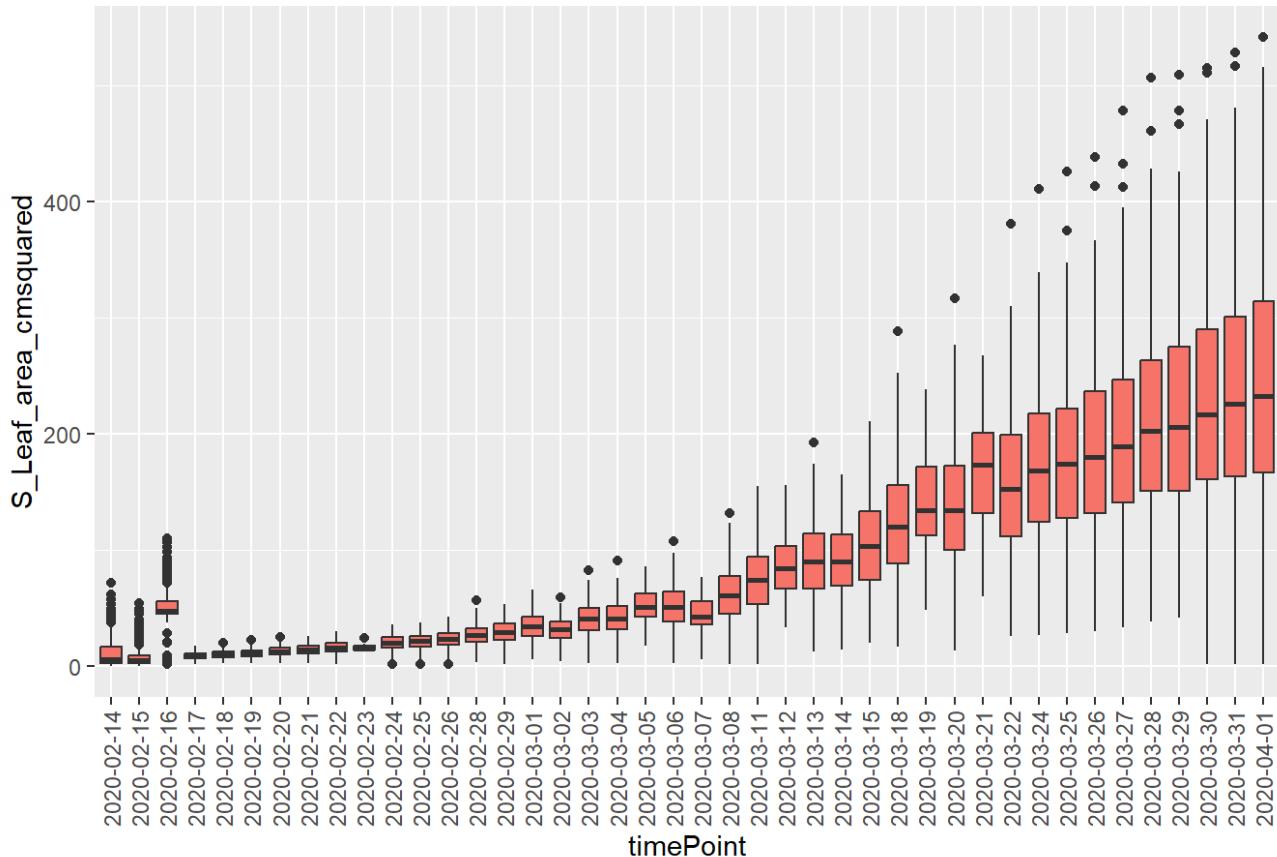
## EPPN2020\_M3P - S\_Leaf\_area\_cmsquared - raw data



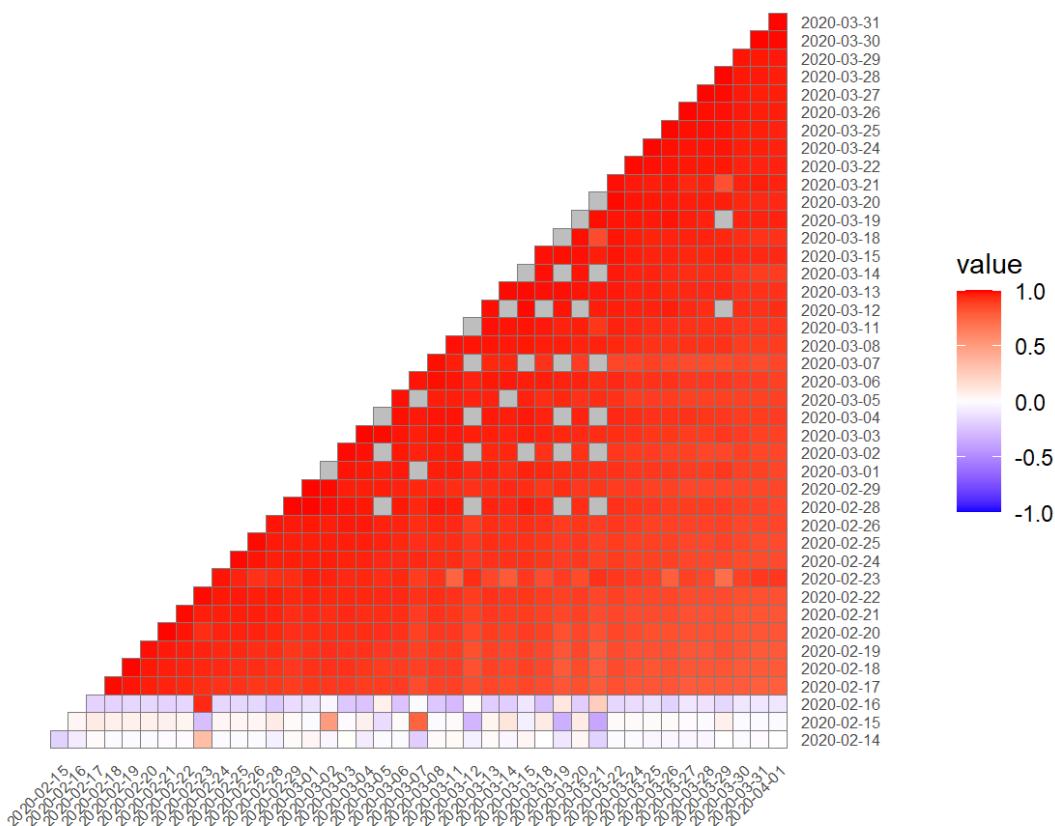
## EPPN2020\_M3P - S\_Leaf\_area\_cmsquared - raw data



## EPPN2020\_M3P - S\_Leaf\_area\_cmsquared



## ?PN2020\_M3P - Correlations of timepoints for S\_Leaf\_area\_cmsquared



## 2. Correction for spatial trends

Fit a model for all time points with no extra fixed effects.

```
#for (trait_name in traits) {
#  single_outliers_name <- paste0("Single_outliers_", trait_name)

#  if (exists(single_outliers_name)) {
#    Single_outliers <- get(single_outliers_name)

#    assign(paste0("modTP_", trait_name),
#           fitModels(TP = Single_outliers,
#                     trait = trait_name,
#                     geno.decomp = "Plant_type"))
#  } else {
#    assign(paste0("modTP_", trait_name),
#           fitModels(TP = timePoint_S,
#                     trait = trait_name,
#                     geno.decomp = "Plant_type"))
#  }
#}
```

print("Erreur dans SpATS::SpATS(response = trait, fixed = fixedForm, random = randForm,  
:

The design matrix associated to the fixed part of the model is not of full rank. Please check that there are no redundant components in the fixed part of the model.")

```
## [1] "Erreur dans SpATS::SpATS(response = trait, fixed = fixedForm, random = randFor  
m, : \n The design matrix associated to the fixed part of the model is not of full ra  
nk. Please check that there are no redundant components in the fixed part of the mode  
l."
```

## Model visualisation

```
for (trait_name in traits) {
  mod_name <- paste0("modTP_", trait_name)

  if (exists(mod_name)) {
    mod <- get(mod_name)

    for (tp in 1:length(num_timepoints$timeNumber)) {
      plot(mod,
            timePoints = tp,
            plotType = "spatial",
            spaTrend = "percentage")
    }

    gif_file <- sprintf("%s/%s_mod.gif", datadir, trait_name)

    plot(mod,
          plotType = "timeLapse",
          spaTrend = "percentage",
          outFile = gif_file)
  } else {
    cat("No model found for", trait_name, "\n")
  }
}
```

```
## No model found for S_Height_cm
## No model found for S_Leaf_area_cmsquared
```

```
for (trait_name in traits) {
  mod_name <- paste0("modTP_", trait_name)
  if (exists(mod_name)) {
    mod <- get(mod_name)

    plot(mod,
          plotType = "rawPred",
          plotLine = TRUE)

    plot(mod,
          plotType = "corrPred",
          plotLine = TRUE)

    plot(mod,
          plotType = "herit",
          yLim = c(0, 0.5))

    plot(mod,
          plotType = "effDim",
          EDTType = "ratio",
          yLim = c(0, 1))

    plot(mod,
          plotType = "corrPred")
  } else {
    cat("No model found for the trait", trait_name, "\n")
  }
}
```

```
## No model found for the trait S_Height_cm
## No model found for the trait S_Leaf_area_cmsquared
```

### 3. Outlier detection for series of observations

By using the splines.

**fitModels**

```
for (trait_name in traits) {
  Spatial_Corrected_name <- paste0("Spatial_Corrected_", trait_name)
  modTP_name <- paste0("modTP_", trait_name)
}
```

```

knots <- c(30)
mintimepoints <- c(9) # Minimal number of observations

for (trait_name in traits) {
  # Nom de la variable pour les données corrigées
  Spatial_Corrected_name <- paste0("Spatial_Corrected_", trait_name)
  modTP_name <- paste0("modTP_", trait_name)

  # Vérifier si le modèle existe
  if (exists(modTP_name)) {
    modTP <- get(modTP_name)
    Spatial_Corrected <- getCorrected(modTP)
    assign(Spatial_Corrected_name, Spatial_Corrected)

    # Ajuster les splines pour les données corrigées
    fit.spline <- fitSpline(inDat = Spatial_Corrected,
                             trait = paste0(trait_name, "_corr"),
                             knots = knots,
                             minNoTP = mintimepoints)

    # Extraire les tables de valeurs prédites et coefficients de splines
    predDat_name <- paste0("predDat_", trait_name)
    coefDat_name <- paste0("coefDat_", trait_name)

    assign(predDat_name, fit.spline$predDat)
    assign(coefDat_name, fit.spline$coefDat)
  } else {
    cat("No model found for", trait_name, "\n")
  }
}

```

```

## No model found for S_Height_cm
## No model found for S_Leaf_area_cmsquared

```

Plot the splines for a plant selection

detectSerieOut

```

thrCor <- c(0.9) # correlation threshold
thrPca <- c(30) # pca angle threshold
thrSlope <- c(0.7) # slope threshold

for (trait_name in traits) {
  Spatial_Corrected_name <- paste0("Spatial_Corrected_", trait_name)
  predDat_name <- paste0("predDat_", trait_name)
  coefDat_name <- paste0("coefDat_", trait_name)

  if (exists(Spatial_Corrected_name) && exists(predDat_name) && exists(coefDat_name)) {
    Spatial_Corrected <- get(Spatial_Corrected_name)
    predDat <- get(predDat_name)
    coefDat <- get(coefDat_name)

    Series_test <- detectSerieOut(corrDat = Spatial_Corrected,
                                    predDat = predDat,
                                    coefDat = coefDat,
                                    trait = paste0(trait_name, "_corr"),
                                    thrCor = thrCor,
                                    thrPca = thrPca,
                                    thrSlope = thrSlope,
                                    geno.decomp = "geno.decomp")

    plot(Series_test, genotypes = levels(factor(Series_test$genotype)))

    assign(paste0("Series_test_", trait_name), Series_test)

    assign(paste0("Spatial_Corrected_Out_", trait_name), Spatial_Corrected)
  } else {
    cat("No corrected data or prediction data found for", trait_name, "\n")
  }
}

```

```

## No corrected data or prediction data found for S_Height_cm
## No corrected data or prediction data found for S_Leaf_area_cmsquared

```

**removeSerieOut**

```

for (trait_name in traits) {
  # Nom de la variable pour les données corrigées
  Spatial_Corrected_name <- paste0("Spatial_Corrected_", trait_name)
  Series_test_name <- paste0("Series_test_", trait_name)

  # Extraire les données corrigées et les résultats des séries
  if (exists(Spatial_Corrected_name) && exists(Series_test_name)) {
    Spatial_Corrected <- get(Spatial_Corrected_name)
    Series_test <- get(Series_test_name)

    # Supprimer les outliers de la série
    Spatial_Corrected_Out <- removeSerieOut(dat = Spatial_Corrected, serieOut = Series_
test)

    # Assigner le résultat à une nouvelle variable
    assign(paste0("Spatial_Corrected_Out_", trait_name), Spatial_Corrected_Out)
  } else {
    cat("No corrected data or series test data found for", trait_name, "\n")
  }
}

```

```

## No corrected data or series test data found for S_Height_cm
## No corrected data or series test data found for S_Leaf_area_cmsquared

```

```

for (trait_name in traits) {
  Spatial_Corrected_Out_name <- paste0("Spatial_Corrected_Out_", trait_name)

  if (exists(Spatial_Corrected_Out_name)) {
    Spatial_Corrected_Out <- get(Spatial_Corrected_Out_name)
    output_file <- sprintf("%s/timeSeriesOutliers_%s.tsv", datadir, trait_name)
    readr::write_tsv(Spatial_Corrected_Out, output_file)

    cat("Data written to:", output_file, "\n")
  } else {
    cat("No corrected data found for", trait_name, "\n")
  }
}

```

```

## No corrected data found for S_Height_cm
## No corrected data found for S_Leaf_area_cmsquared

```

## 4. With the cleaned data, re-do the spatial correction

This is used to compare the values before and after.

Need to write a for loop for all the variables.

For S\_Height\_cm

Estimation of parameter from time series

For S\_Area\_cmsquared

Estimation of parameter from time series