

# ST117 FINAL Written Report - Phase 2 Task B

Report Pod 041

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## Contributions

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## Setting up data frames before for phase 2 part B

```
#Import the dataframes from phase 1
df_stream_wide <- readRDS("/Users/danielguo/Desktop/University/Year 1/ST117/5645242_041_WR/df_stream_wide.rds")
df_precipitation_wide <- readRDS("/Users/danielguo/Desktop/University/Year 1/ST117/5645242_041_WR/df_precipitation_wide.rds")
df_soil_wide <- readRDS("/Users/danielguo/Desktop/University/Year 1/ST117/5645242_041_WR/df_soil_wide.rds")

#Averaging the repeated sampling by the variable LCODE
df_stream_avg <- df_stream_wide %>%
  group_by(SDATE, SITECODE, LCODE) %>%
  summarise(across(where(is.numeric), \(x) mean(x, na.rm = TRUE)), .groups = "drop")

#weighted averaging considering the sampling volume (VOLUME) by the variable RID
df_soil_avg <- df_soil_wide %>%
  group_by(SDATE, SITECODE, RID) %>%
  summarise(across(where(is.numeric), \(x) weighted.mean(x, w = VOLUME, na.rm = TRUE)), .groups = "drop")

#STAGE, VACUUM, and VOLUME are only contained in some of the datasets and may be dropped unless needed.
df_stream_B <- df_stream_avg %>%
  select(-STAGE)
df_precipitation_B <- df_precipitation_wide %>%
  select(-VOLUME)
df_soil_B <- df_soil_avg %>%
  select(-VACUUM, -VOLUME)
```

## Question 1 and 2

```
# Function to convert the dataframes into averaged data by weeks for each sites
convert_to_weekly <- function(df, date_col = "SDATE", site_col = "SITECODE") {
  df <- df %>%
    mutate(
      DATE = as.Date(.data[[date_col]]), #Convert date column to Date type
      YEAR = format(DATE, "%Y"), # Extract year
      WEEK = format(DATE, "%W") # Extract week
    ) %>%
    group_by(.data[[site_col]], YEAR, WEEK) %>% # Group by different site, year, and week
```

```

    summarise(across(where(is.numeric), ~ mean(.x, na.rm = TRUE)), .groups = "drop") #Average numeric columns b

  return(df)
}

# function to merge three datasets (stream, precipitation, soil)
merge_datasets <- function(stream, precipitation, soil) {
  # Rename columns to Variable_dataset abbreviations
  stream <- stream %>% rename_with(~ paste0(., "_str"), -c(SITECODE, YEAR, WEEK))
  precipitation <- precipitation %>% rename_with(~ paste0(., "_prec"), -c(SITECODE, YEAR, WEEK))
  soil <- soil %>% rename_with(~ paste0(., "_soil"), -c(SITECODE, YEAR, WEEK))
  # merge and full_join all the datasets
  merged <- full_join(stream, precipitation, by = c("SITECODE", "YEAR", "WEEK")) %>%
    full_join(soil, by = c("SITECODE", "YEAR", "WEEK"))

  return(merged)
}

# Function to remove outliers from the applicable numeric columns using interquartile range formula
remove_outliers <- function(df) {
  df %>% mutate(across(where(is.numeric), ~ {
    Q1 <- quantile(.x, 0.25, na.rm = TRUE)
    Q3 <- quantile(.x, 0.75, na.rm = TRUE)
    IQR <- Q3 - Q1
    low <- Q1 - 1.5 * IQR
    high <- Q3 + 1.5 * IQR
    ifelse(.x < low | .x > high, NA, .x)
  })))
}

# function to compute correlation matrix of one chemical between different datasets for given site and year range
variable_correlation <- function(merged_df, variable, site, year_range, method = "pearson") { #setting pearson a
  df <- merged_df %>%
    filter(SITECODE == site, as.numeric(YEAR) %in% year_range) %>%
    select(matches(paste0("^", variable, "_"))) %>%
    na.omit()

  if (nrow(df) < 2) {
    stop("Not enough data for correlation.") #halt if insufficient data
  }

  if (sum(complete.cases(df)) / nrow(df) < 0.5) {
    warning("More than 50% of data is missing for this site/variable") # Warning if there exist many NAs
  }

  # Use cat instead of message to prevent double output
  cat("Correlation matrix for", variable,
      "from stream, precipitation, and soil solution (Site:", site,
      "; Years:", paste(range(year_range), collapse = "-"), ")\n")

  round(cor(df, method = method), 2) #rounded correlation matrix to 2 decimal places
}

# function to summarise the correlation data in a table formate
summary_correlation_across_sites <- function(merged_df, variable, sites, year_range, method = "pearson") {
  result <- lapply(sites, function(site) {
    tryCatch({
      corr <- variable_correlation(merged_df, variable, site, year_range, method) # computing correlation
      data.frame(Site = site,
                  Str_vs_Prec = corr[1, 2],
                  Str_vs_Soil = corr[1, 3],

```

```

        Prec_vs_Soil = corr[2, 3],
        stringsAsFactors = FALSE)
}, error = function(e) { #if fail, return NA
  data.frame(Site = site,
             Str_vs_Prec = NA,
             Str_vs_Soil = NA,
             Prec_vs_Soil = NA,
             stringsAsFactors = FALSE)
})
})

do.call(rbind, result) #combining all the results into one dataframe
}

#a function to plot the correlation summary as a heatmap for visualisatio.
plot_correlation_summary <- function(summary_df, variable) {
  summary_df_long <- summary_df %>%
    pivot_longer(cols = -Site, names_to = "Comparison", values_to = "Correlation") #reshape the dataframe for easier use

  ggplot(summary_df_long, aes(x = Comparison, y = Site, fill = Correlation)) +
    geom_tile() + # coloured tile heatmap
    geom_text(aes(label = round(Correlation, 2)), size = 3) + # add correlation coefficient over the heatmap for easier use
    scale_fill_gradient2(low = "darkblue", high = "maroon", #colour coding
                        midpoint = 0, limit = c(-1, 1)) +

  theme_minimal() +
  labs(
    title = paste("Correlation Summary for", variable, "Across Sites"),
    x = "Comparison", y = "Site"
  ) +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1)) # rotate x-axis labels for better readability
}

```

## Question 3

### 1. convert\_to\_weekly

**Purpose:** Converts any water chemistry dataset into a weekly data by averaging repeated measurements within the week.

**Arguments:**

- df (data.frame): Input dataframes (e.g., df\_stream\_B, df\_precipitation\_B, df\_soil\_B).
- date\_col (character): Column name for date variable (default: "SDATE": no need to input)
- site\_col (character): Column name for site code variable (default: "SITECODE": no need to input)

**Returns:** A data frame sorted by site, year, and week, with averaged numeric variables for each weeks.

```

# Example: (set the data into another dataframe for easier use in later functions)
df_stream_weekly <- convert_to_weekly(df_stream_B)
df_precipitation_weekly <- convert_to_weekly(df_precipitation_B)
df_soil_weekly <- convert_to_weekly(df_soil_B)
head(df_stream_weekly)

```

```

## # A tibble: 6 x 23
##   SITECODE YEAR  WEEK  LCODE ALUMINIUM TOTALN CHLORIDE  DOC  IRON MAGNESIUM
##   <chr>    <chr> <chr> <dbl>    <dbl>    <dbl>    <dbl> <dbl> <dbl>    <dbl>
## 1 T02     1993  19     1      0.053    NaN     6.89   6.8   0.14    1.38
## 2 T02     1993  20     1      NaN      NaN     NaN     NaN   NaN     NaN
## 3 T02     1993  21     1      NaN      NaN     NaN     NaN   NaN     NaN
## 4 T02     1993  22     1      0.282    NaN     5.84  13.3   0.369    1.02

```

```
## 5 T02      1993  23      1 NaN      NaN      NaN      NaN      NaN
## 6 T02      1993  24      1 NaN      NaN      NaN      NaN      NaN
## # i 13 more variables: NH4N <dbl>, NO3N <dbl>, PH <dbl>, PO4P <dbl>,
## #   POTASSIUM <dbl>, SO4S <dbl>, SODIUM <dbl>, CALCIUM <dbl>, PHAQCS <dbl>,
## #   PHAQCU <dbl>, ALKY <dbl>, CONDY <dbl>, TOTALP <dbl>
```

```
head(df_precipitation_weekly)
```

```
## # A tibble: 6 x 22
##   SITECODE YEAR  WEEK ALUMINIUM SODIUM  SO4S POTASSIUM PO4P    PH NO3N  NH4N
##   <chr>    <chr> <chr>    <dbl>  <dbl> <dbl>    <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 T01      1992  50      NaN    0    0      0    NaN  5.3    0    1.7
## 2 T01      1993  01      NaN    0    0      0    NaN  5.2    0.5  1.9
## 3 T01      1993  02      0    0    0      0    0    5.3    0.3  0.4
## 4 T01      1993  03      0    0  15.5      0    0    5      0    0.5
## 5 T01      1993  04      0    0  41.3      0    0    5.1    0.7  0.6
## 6 T01      1993  05      NaN    NaN NaN      NaN    NaN  4.6    0.7  1
## # i 11 more variables: MAGNESIUM <dbl>, IRON <dbl>, DOC <dbl>, CHLORIDE <dbl>,
## #   CALCIUM <dbl>, TOTALN <dbl>, PHAQCS <dbl>, PHAQCU <dbl>, CONDY <dbl>,
## #   ALKY <dbl>, TOTALP <dbl>
```

```
head(df_soil_weekly)
```

```
## # A tibble: 6 x 22
##   SITECODE YEAR  WEEK  SO4S SODIUM TOTALN ALUMINIUM CALCIUM CHLORIDE  DOC
##   <chr>    <chr> <chr> <dbl>  <dbl>  <dbl>    <dbl>    <dbl>    <dbl> <dbl>
## 1 T01      1993  15      NaN    NaN    NaN      NaN      NaN      NaN    NaN
## 2 T01      1993  43      NaN    NaN    NaN      NaN      NaN      NaN    NaN
## 3 T01      1993  45      NaN    NaN    NaN      NaN      NaN      NaN    NaN
## 4 T01      1993  47      NaN    NaN    NaN      NaN      NaN      NaN    NaN
## 5 T01      1993  49      NaN    NaN    NaN      NaN      NaN      NaN    NaN
## 6 T01      1993  51      NaN    NaN    NaN      NaN      NaN      NaN    NaN
## # i 12 more variables: IRON <dbl>, MAGNESIUM <dbl>, NH4N <dbl>, NO3N <dbl>,
## #   PH <dbl>, PO4P <dbl>, POTASSIUM <dbl>, PHAQCS <dbl>, PHAQCU <dbl>,
## #   ALKY <dbl>, CONDY <dbl>, TOTALP <dbl>
```

## 2. merge\_datasets

**Purpose:** Combines stream water, precipitation, and soil solution dataframes into a large merged dataframe.

### Arguments:

- stream (data.frame): Weekly stream dataset. (df\_stream\_weekly)
- precipitation (data.frame): Weekly precipitation dataset.(df\_precipitation\_weekly)
- soil(data.frame): Weeklysoil dataset.(df\_soil\_weekly)

(df\_stream\_weekly, df\_precipitation\_weekly, df\_soil\_weeklyfrom last function's example)

**Returns:** A data.frame merged by SITECODE, YEAR, and WEEK, with columns named like ALUMINIUM\_str, ALUMINIUM\_prec, ALUMINIUM\_soil

```
# Example: (set the data into another dataframe for easier use in later functions)
df_merged <- merge_datasets(df_stream_weekly, df_precipitation_weekly, df_soil_weekly)
head(df_merged)
```

```
## # A tibble: 6 x 61
##   SITECODE YEAR  WEEK LCODE_str ALUMINIUM_str TOTALN_str CHLORIDE_str DOC_str
##   <chr>    <chr> <chr>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>
## 1 T02      1993  19      1      0.053      NaN      6.89      6.8
## 2 T02      1993  20      1      NaN      NaN      NaN      NaN
## 3 T02      1993  21      1      NaN      NaN      NaN      NaN
## 4 T02      1993  22      1      0.282      NaN      5.84     13.3
## 5 T02      1993  23      1      NaN      NaN      NaN      NaN
```

```
## 6 T02      1993  24      1      NaN      NaN      NaN      NaN
## # i 53 more variables: IRON_str <dbl>, MAGNESIUM_str <dbl>, NH4N_str <dbl>,
## #   NO3N_str <dbl>, PH_str <dbl>, PO4P_str <dbl>, POTASSIUM_str <dbl>,
## #   SO4S_str <dbl>, SODIUM_str <dbl>, CALCIUM_str <dbl>, PHAQCS_str <dbl>,
## #   PHAQCUC_str <dbl>, ALKY_str <dbl>, CONDY_str <dbl>, TOTALP_str <dbl>,
## #   ALUMINIUM_prec <dbl>, SODIUM_prec <dbl>, SO4S_prec <dbl>,
## #   POTASSIUM_prec <dbl>, PO4P_prec <dbl>, PH_prec <dbl>, NO3N_prec <dbl>,
## #   NH4N_prec <dbl>, MAGNESIUM_prec <dbl>, IRON_prec <dbl>, DOC_prec <dbl>, ...
```

### 3. remove\_outliers

**Purpose:** Removes outliers from the dataframe using the interquartile range method. Values outside  $[Q1 - 1.5 \times IQR, Q3 + 1.5 \times IQR]$  are replaced with NA.

#### Arguments:

- `df_merged` (data.frame): Merged dataset. (`df_merged` from last function, `merge_datasets()`, example)

**Returns:** A data.frame with outlier values replaced by NA.

```
# Example: (set the data into another dataframe for easier use in later functions)
df_merged_no_outliers <- remove_outliers(df_merged)
head(df_merged_no_outliers)
```

```
## # A tibble: 6 x 61
##   SITECODE YEAR  WEEK  LCODE_str ALUMINIUM_str TOTALN_str CHLORIDE_str DOC_str
##   <chr>      <chr> <chr>      <dbl>          <dbl>        <dbl>        <dbl>    <dbl>
## 1 T02      1993  19      1          0.053          NA          6.89      6.8
## 2 T02      1993  20      1          NA            NA          NA        NA
## 3 T02      1993  21      1          NA            NA          NA        NA
## 4 T02      1993  22      1          NA            NA          5.84     13.3
## 5 T02      1993  23      1          NA            NA          NA        NA
## 6 T02      1993  24      1          NA            NA          NA        NA
## # i 53 more variables: IRON_str <dbl>, MAGNESIUM_str <dbl>, NH4N_str <dbl>,
## #   NO3N_str <dbl>, PH_str <dbl>, PO4P_str <dbl>, POTASSIUM_str <dbl>,
## #   SO4S_str <dbl>, SODIUM_str <dbl>, CALCIUM_str <dbl>, PHAQCS_str <dbl>,
## #   PHAQCUC_str <dbl>, ALKY_str <dbl>, CONDY_str <dbl>, TOTALP_str <dbl>,
## #   ALUMINIUM_prec <dbl>, SODIUM_prec <dbl>, SO4S_prec <dbl>,
## #   POTASSIUM_prec <dbl>, PO4P_prec <dbl>, PH_prec <dbl>, NO3N_prec <dbl>,
## #   NH4N_prec <dbl>, MAGNESIUM_prec <dbl>, IRON_prec <dbl>, DOC_prec <dbl>, ...
```

### 4. variable\_correlation

**Purpose:** Generate a correlation matrix for a given variable across the three water types for a given site and year range.

#### Arguments:

- `df_merged_no_outliers` (data.frame): Cleaned merged dataset.
- `variable` (character): Name of the variable ("ALUMINIUM").
- `site` (character): Site code ("T02").
- `year_range` (numeric vector): Range of years to include (1992:1993). (data from 1992 to 2005)
- `method` (character): "pearson" (default) or "spearman". (no need to input)

**Returns:** A 3×3 correlation matrix between the variable's values in `_str`, `_prec`, and `_soil`.

```
#For example: Aluminuim at site T02 for 1992-1993 (do not set to dataframe)
variable_correlation(df_merged_no_outliers, "ALUMINIUM", "T02", 1992:1993)
```

```
## Correlation matrix for ALUMINIUM from stream, precipitation, and soil solution (Site: T02 ; Years: 1992-1993)
```

```
##           ALUMINIUM_str ALUMINIUM_prec ALUMINIUM_soil
## ALUMINIUM_str           1.00          -0.59          -0.75
## ALUMINIUM_prec          -0.59           1.00          -0.07
## ALUMINIUM_soil          -0.75          -0.07           1.00
```

## 5. summary\_correlation\_across\_sites

**Purpose:** Generates a summary table of pairwise correlations for a given variable across different data, for a given site and given year range.

### Arguments:

- `df_merged_no_outliers` (data.frame): Cleaned merged dataset.
- `variable` (character): Name of the variable (“ALUMINIUM”).
- `site` (character): Site code (“T02”).
- `year_range` (numeric vector): Range of years to include (1992:1993). (data from 1992 to 2005)
- `method` (character): “pearson” (default) or “spearman”. (don’t need to input)

**Returns:** A data frame with correlations between water types.

*#For example: Aluminuim at site T02 for 1992-1993 (set the data into another dataframe for easier use in the la*  
`summary_alum_T02_9293 <- summary_correlation_across_sites(df_merged_no_outliers, "ALUMINIUM", "T02", 1992:1993)`

```
## Correlation matrix for ALUMINIUM from stream, precipitation, and soil solution (Site: T02 ; Years: 1992-1993)
summary_alum_T02_9293
```

```
##   Site Str_vs_Prec Str_vs_Soil Prec_vs_Soil
## 1  T02          -0.59        -0.75        -0.07
```

## 6. plot\_correlation\_summary

**Purpose:** Generates a heatmap that provides visualisation for a given variable for a given year range.

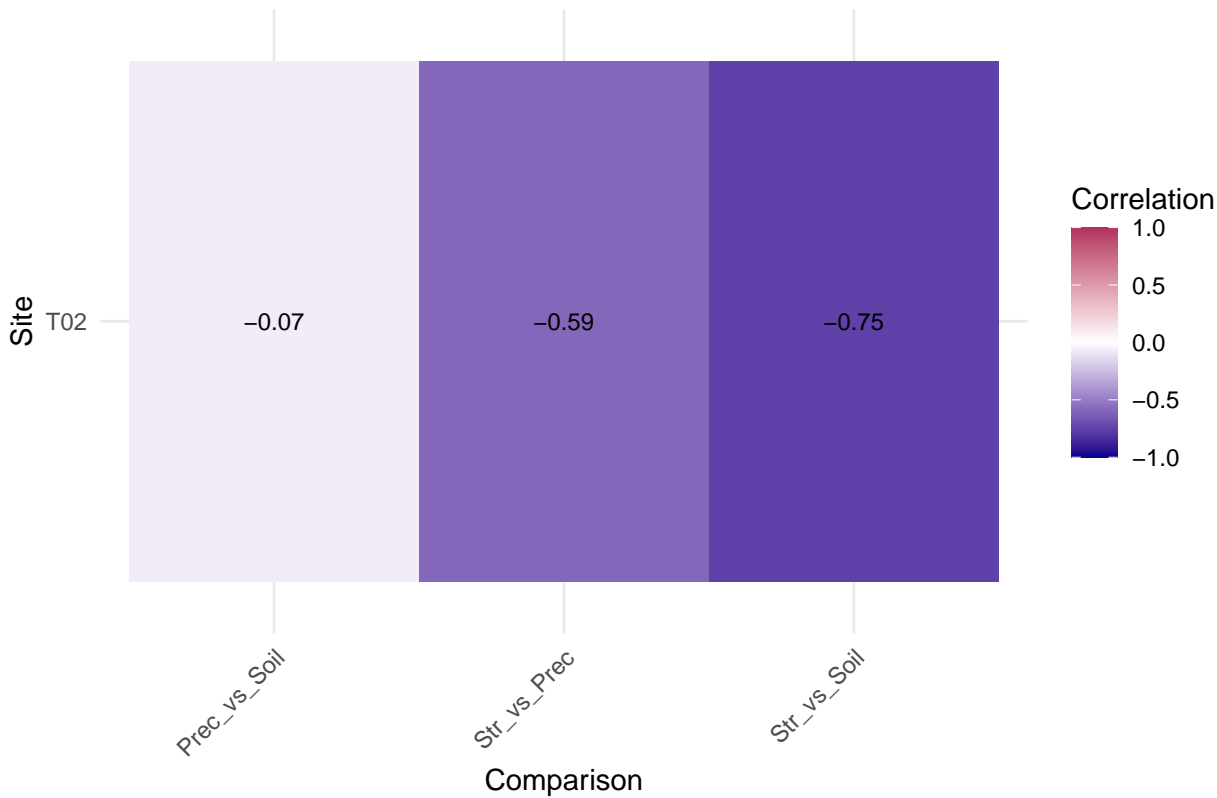
### Arguments:

- `summary_df` (summary\_alum\_T02\_9293 from last function’s example) (data.frame): Summary with site-wise correlations of an varianle between water types.
- `variable` (character): (Aluminium for example): Name of the variable (e.g., “ALUMINIUM”).

**Returns:** A data frame with site-wise correlations between water types for a given variablea nd year range.

*#For example: Aluminuim at site T02 for 1992-1993*  
`plot_correlation_summary(summary_alum_T02_9293, "ALUMINIUM")`

## Correlation Summary for ALUMINIUM Across Sites



## Question 4

### Hypotheses 1:

```
sites_1 <- c("T02", "T04", "T05", "T06", "T07", "T11") # sites we want to examine
summary_correlation_across_sites(df_merged_no_outliers, variable = "POTASSIUM", sites = sites_1, 2010:2015, method = "pearson")

## Correlation matrix for POTASSIUM from stream, precipitation, and soil solution (Site: T02 ; Years: 2010-2015)
## Correlation matrix for POTASSIUM from stream, precipitation, and soil solution (Site: T04 ; Years: 2010-2015)
## Correlation matrix for POTASSIUM from stream, precipitation, and soil solution (Site: T05 ; Years: 2010-2015)
## Correlation matrix for POTASSIUM from stream, precipitation, and soil solution (Site: T06 ; Years: 2010-2015)
## Correlation matrix for POTASSIUM from stream, precipitation, and soil solution (Site: T07 ; Years: 2010-2015)
## Correlation matrix for POTASSIUM from stream, precipitation, and soil solution (Site: T11 ; Years: 2010-2015)

##   Site Str_vs_Prec Str_vs_Soil Prec_vs_Soil
## 1  T02      0.56      0.32      0.35
## 2  T04      0.35      0.22      0.26
## 3  T05      0.16      0.34      0.16
## 4  T06      0.38      0.38     -0.17
## 5  T07      0.16      0.18      0.27
## 6  T11      0.29      0.32      0.01
```

The abundance of Potassium (essential for fertile land) in soil solution is weakly positively correlated with stream water. This is supported by Str\_vs\_Soil Pearson correlations consistently falling within [0.15, 0.5] between 2010–2015 across sites T02, T07, T11, and T12, as shown in the summary table.

### Hypotheses 2:

```
sites_2 <- c("T02", "T04", "T05", "T07", "T12")
summary_correlation_across_sites(df_merged_no_outliers, variable = "NO3N", sites = sites_2, 1992:2015, method = "pearson")

## Correlation matrix for NO3N from stream, precipitation, and soil solution (Site: T02 ; Years: 1992-2015 )
## Correlation matrix for NO3N from stream, precipitation, and soil solution (Site: T04 ; Years: 1992-2015 )
## Correlation matrix for NO3N from stream, precipitation, and soil solution (Site: T05 ; Years: 1992-2015 )
```

## Correlation matrix for NO3N from stream, precipitation, and soil solution (Site: T07 ; Years: 1992-2015 )  
 ## Correlation matrix for NO3N from stream, precipitation, and soil solution (Site: T12 ; Years: 1992-2015 )

##	Site	Str_vs_Prec	Str_vs_Soil	Prec_vs_Soil
## 1	T02	-0.04	0.20	-0.01
## 2	T04	0.25	0.31	0.13
## 3	T05	0.05	0.22	0.08
## 4	T07	-0.07	0.34	0.06
## 5	T12	0.26	0.61	0.16

There is no relationship of the abundance of Nitrate Oxygen (abundant in acid rain) in stream water and precipitation. The absolute values of the Str\_vs\_Prec Spearman correlations exceed 0.2 only twice during the full study period, as shown in the summary table across selected sites.