Sensitivity of CO2 emission rate to temperature in cranberry soils

Dr. Wilfried Dossou-Yovo

13/02/2022

Wilfried Dossou-Yovo, Serge Étienne Parent, Noura Ziadi, Bertrand Normand, and Léon Étienne parent print(Sys.Date())

[1] "2022-10-10"

1. Objective

This notebook generate the result of CO2 data analysis. Data set contains a collection of soil caracteristics, measured co2 emission collected from incubation study. Soil samples was collected from two cranberry fied stand of eastern canada. Incubation study was carried out at Agriculture and Agri-food Canada(sainte-foy, quebec,qc) from February to Mai 2019. The aim of this study was to measure CO2 emission rates in cranberry soils of Eastern Canada as related to soil temperature and depth

2. Statistical questions

In addition to data exploration, this notebook will answer the following statistical questions.

- 1. What is the influence of soil depth and temperature on CO2 emission?
- 2. Can Arrhenius equation and Q10 be useful to describe temperature sensitivity of carbon decomposition across layers?

3. Packages

We need package tidyverse which loads a set of packages for easy data manipulation(Ex: dplyr) and visualization (ex: ggplot2). We also use ggpubr to customise publication ready plot, ggpmisc and grid are useful packages as extensions to ggplot2.

4. Import data

We load two data_pot and data_co2 involved in our anylisis. data_pot contained details about sites sampling, soil sampling(soil depth, weight, water content and bulk density), laboratory incubation temperature while data_co2 contained details about laboratory incubation time, co2 emission and jar masson details. data_co2 was combined with data_pot with left_join function

ID potSites	Depth (cm) Rep	etitionTen	nperaturePo (°C)	ot weight So (g)	oil weight (g)	Water volumecor (ml)	Water ntent (%)	Bulk density (g/cm3)
6A9	10	1	10	244.81	110.32	42.1	10.3	0.89
21A9	10	1	20	251.14	110.33	42.1	10.3	0.89
54A9	10	1	30	249.63	110.33	42.1	10.3	0.89

ID potSites	Depth (cm) Rep	$\operatorname{etition} \operatorname{Tem}$	peraturePo (°C)	ot weight So (g)	oil weight (g)	Water volumecon (ml)	Water atent (%)	Bulk density (g/cm3)
18A9	10	2	10	246.04	124.93	27.5	24.9	0.89
59A9	10	2	20	248.21	124.93	27.5	24.9	0.89
60A9	10	2	30	255.03	124.90	27.5	24.9	0.89

5. Some calculations

Several variables have been added to our data in order to proceed for analysis. The added variables are the following: Temperature (Kelvin), Molar Volume (L/mol), Headspace Volume (mL), Dry soil weight (g), CO2 emission (ug/h/g), CO2 emission (mg/kg), decomposition rate K, lnKand 1/T(T = Temperature(Kelvin))

6. Exploratory data analysis

6.1 Histogram

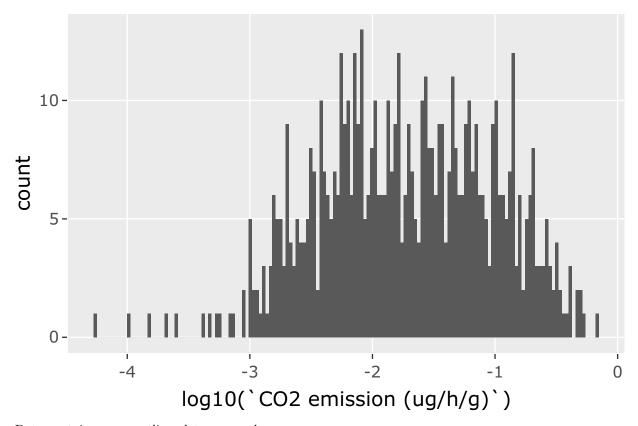
```
New.labs <- c("10°C", "20°C", "30°C") # Change labels
names(New.labs) <- c("10", "20", "30")

New.labs_b <- c("[0-10 cm]", "[10-20 cm]", "[20-30 cm]") # Change labels
names(New.labs_b) <- c("10", "20", "30")

library(plotly)
ggplotly(
   data_co2 |>
        ggplot() +
        geom_histogram(aes(x = log10(`CO2 emission (ug/h/g)`)), bins = 150)
)

## Warning in FUN(X[[i]], ...): NaNs produced

## Warning: Removed 37 rows containing non-finite values (stat_bin).
```



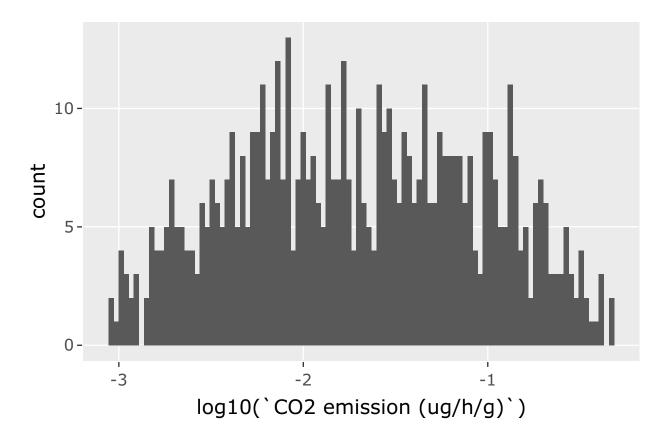
Data contains some outliers, let remove them

```
data_co2_clean <- data_co2 |>
  mutate(log_tr = log10(`CO2 emission (ug/h/g)`)) |>
  filter(log_tr > -3.06 & log_tr < -0.33) |>
  drop_na()
```

Warning in mask\$eval_all_mutate(quo): NaNs produced

Now data look well distributed

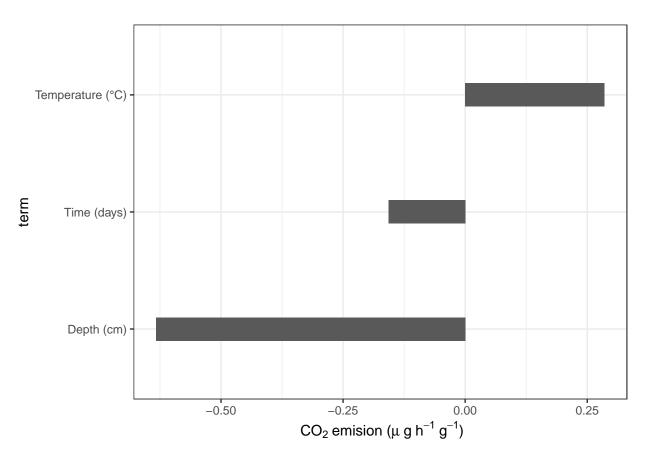
```
ggplotly(
  data_co2_clean |>
    ggplot() +
    geom_histogram(aes(x = log10(`CO2 emission (ug/h/g)`)), bins = 100)
)
```



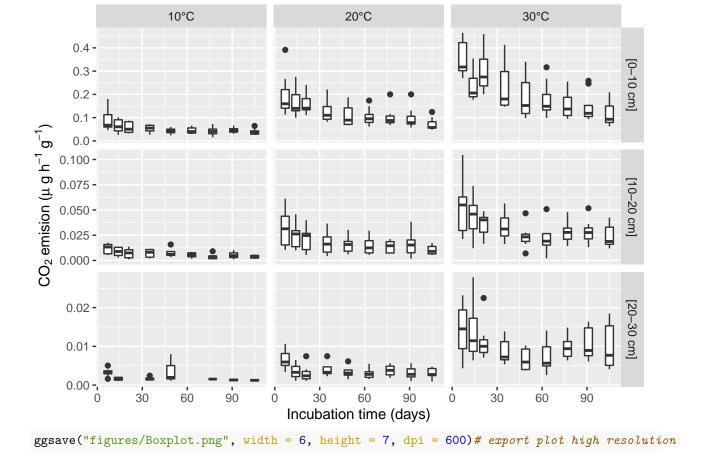
6.2 Correlations

```
data_co2_clean |>
  select(`Time (days)`, `Depth (cm)`, `Temperature (°C)`, `CO2 emission (ug/h/g)`) |>
  corrr::correlate() |>
  corrr::focus(`CO2 emission (ug/h/g)`) |>
  mutate(term = fct_reorder(term, `CO2 emission (ug/h/g)`)) |>
  ggplot(aes(x = `CO2 emission (ug/h/g)`, y= term)) +
  geom_col(width = 0.2) +
  labs(x = bquote(~CO[2]~ 'emision ('*mu~'g'~ h^-1~g^-1*')')) +
  theme_bw()
```

##
Correlation method: 'pearson'
Missing treated using: 'pairwise.complete.obs'



6.3 Boxplot



7. What is the influence of soil depth and temperature on CO2 emission?

7.1 Build model: linear regression

```
model_fit <- model_spec |>
fit(`CO2 emission (ug/h/g)` ~ ., data_co2_preprocessed)
```

Fit model

```
tidy(model_fit)
Exploring model results
## # A tibble: 5 x 5
##
     term
                           estimate std.error statistic
                                                              p.value
##
     <chr>
                                         <dbl>
                                                                <dbl>
                              <dbl>
                                                     <dbl>
## 1 (Intercept)
                                       0.00983
                            -1.70
                                                   -173.
## 2 `Time (days)`
                                       0.00985
                                                    -10.5 1.22e- 23
                            -0.103
## 3 `Depth (cm)`
                            -0.579
                                       0.00993
                                                    -58.3 3.08e-247
## 4 `Temperature (°C)`
                                       0.00992
                             0.273
                                                     27.6 3.81e-108
## 5 Sites_PF45
                            -0.0202
                                       0.00984
                                                     -2.05 4.11e- 2
glance(model_fit)
## # A tibble: 1 x 12
##
     r.squared adj.r.squared sigma statistic
                                                     p.value
                                                                 df logLik
                                                                               AIC
                                                                                      BIC
##
          <dbl>
                         <dbl> <dbl>
                                                                      <dbl> <dbl> <dbl>
                                            <dbl>
                                                       <dbl> <dbl>
          0.868
                         0.867 0.240
                                             969. 6.71e-258
                                                                       7.36 -2.71 23.6
                                                                  4
## # ... with 3 more variables: deviance <dbl>, df.residual <int>, nobs <int>
par(mfrow=c(2,2)) # plot all 4 plots in one
plot(model_fit$fit,
     pch = 16,
     col = '#006EA1')
                                                                        Standardized residuals
                                                                                            Normal Q-Q
                                      Residuals vs Fitted
                          2
                     Residuals
                          o.
                                                                             0
                          -1.0
                              -3.0
                                    -2.5
                                           -2.0
                                                  -1.5
                                                        -1.0
                                                               -0.5
                                                                                                             2
                                                                                                                  3
                                                                                   -3
                                          Fitted values
                                                                                         Theoretical Quantiles
                     (Standardized residuals)
                                                                        Standardized residuals
                                        Scale-Location
                                                                                      Residuals vs Leverage
                                                                             ^{\circ}
                                                                             7
                                                                             တု
```

Inspect the model

-2.0 -1.5

Fitted values

-1.0

-0.5

0.000

0.005

Leverage

0.010

0.01

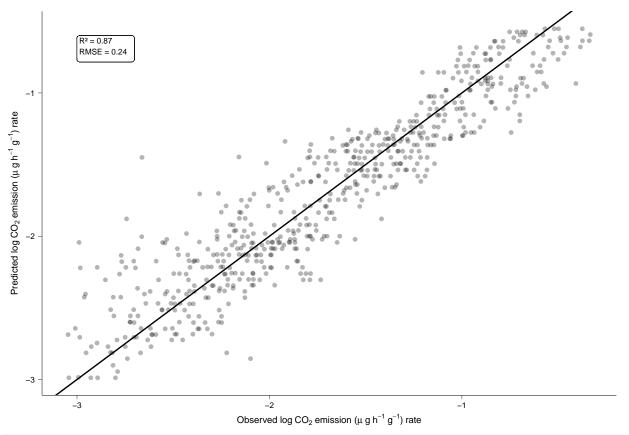
-3.0

-2.5

```
prediction <- model_fit |>
  predict(data_co2_preprocessed)
```

Prediction

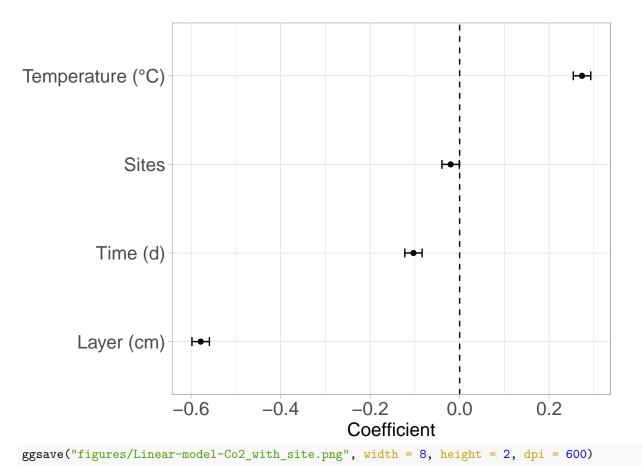
```
rmse <- data_co2_preprocessed |>
     bind_cols(prediction) |>
     rmse(`CO2 emission (ug/h/g)`, .pred)
collect Metrics
## # A tibble: 1 x 3
              .metric .estimator .estimate
##
              <chr> <chr>
                                                                               <dh1>
## 1 rmse
                                  standard
                                                                               0.239
rmse <- round(as.numeric(rmse[1,3]), 2)</pre>
rsq <- data_co2_preprocessed |>
    bind_cols(prediction) |>
     rsq(`CO2 emission (ug/h/g)`, .pred)
rsq
## # A tibble: 1 x 3
              .metric .estimator .estimate
              <chr> <chr>
## 1 rsq
                                    standard
                                                                               0.868
rsq <- round(as.numeric(rsq[1,3]), 2)</pre>
options(repr.plot.width = 4, repr.plot.height = 2)
px <- data_co2_preprocessed |>
     bind_cols(prediction) |>
      ggplot(aes(x = CO2 emission (ug/h/g), y = .pred)) +
      geom_point(size = 1, alpha = .3) +
      geom_label(aes(x = -3, y = -.6),
                                    vjust = 1, hjust = 0, size = 2, label.size = 0.1,
                                    label = paste("R2 =", rsq, "\nRMSE =", rmse)) +
      geom_abline() +
      theme_pubr() +
      theme(axis.title=element_text(size=7),
                      axis.line = element_line(size = 0.1),
                      axis.ticks = element_line(size = 0.1),
                      axis.text = element_text(size = 6)) +
      labs(x=bquote("Observed log"~CO[2]~'emission ('*mu~'g'~h^-1~g^-1*')'~"rate")~,~y=bquote("Predict log"~CO[2]~'emission ('*mu~'g'~h^-1~g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'~h^-1*'g'
рх
```



ggsave("figures/Observed and predicted co2 emission.png", width = 4, height = 2.2, dpi = 600)

7.3 Variable coefficient and confidence intervals

```
options(repr.plot.width = 8, repr.plot.height = 2)
term_rename <- tibble(term = c("`Time (days)`", "`Depth (cm)`", "`Temperature (°C)`", "Sites_PF45"),</pre>
                      name_corrected = c("Time (d)", "Layer (cm)", "Temperature (°C)", "Sites"))
h <- broom::tidy(model_fit, conf.int = TRUE) |>
  dplyr::filter(term != "(Intercept)") |>
  left_join(term_rename, by = "term") |>
  mutate(term_rename = fct_reorder(name_corrected, estimate)) |>
  ggplot(aes(estimate, term rename)) +
  geom_vline(xintercept = 0, linetype = 2) +
  geom_point() +
  geom_errorbarh(aes(xmin = conf.low, xmax = conf.high), height = 0.1,
                 size=0.5) +
  labs(x = "Coefficient", y = "") +
  theme_light() +
  theme(axis.text=element_text(size=14),
        axis.title=element_text(size=14)) # Time (d)
```

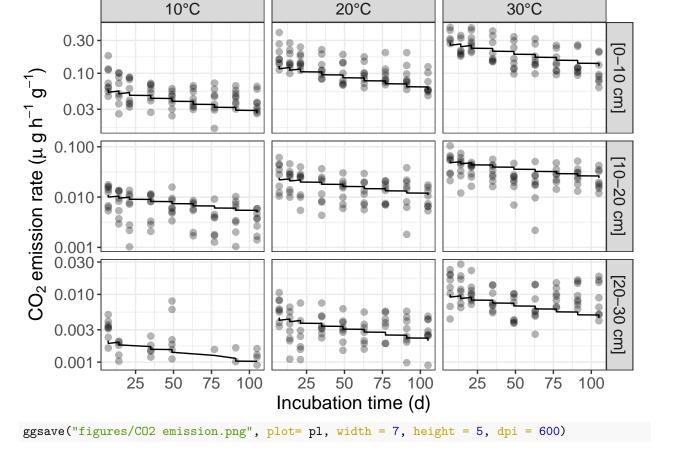


7.3 Prediction model of CO2 emission in cranberry soils in three-layer positions (0-10 cm, 10-20

cm, 20-30 cm) and at three temperatures (10, 20 and 30oC).

pl

```
New.labs <- c("10°C", "20°C", "30°C") # Change labels
names(New.labs) <- c("10", "20", "30")
New.labs_b \leftarrow c("[0-10 cm]", "[10-20 cm]", "[20-30 cm]") # Change labels
names(New.labs b) \leftarrow c("10", "20", "30")
options(repr.plot.width = 8, repr.plot.height = 6)
pl <- data_co2_clean |>
  bind_cols(10^prediction) |>
  ggplot(aes(x = Time (days)), y = CO2 emission (ug/h/g))) +
  geom_point(size = 2, alpha = .3) +
  facet_grid(`Depth (cm)` ~ `Temperature (°C)`, scales = "free", labeller = labeller(`Depth (cm)` = New
  geom_line(aes(x = `Time (days)`, y = `.pred`)) +
  scale_y_log10() +
  theme bw() +
  theme(strip.text = element_text(size = 12), axis.text=element_text(size=12),
       axis.title=element_text(size=14),
        axis.title.y = element_text(size=14)) +
  xlab("Incubation time (d)") + ylab(bquote(~CO[2]~ 'emission rate ('*mu~'g'~ h^-1~g^-1*')'))
```



8. What is the temperature sensitivity across cranberry soil layers?

8.1 Fit of Arrhenius equation

The Arrhenius equation has been used to describe temperature sensitivity to CO2 emission. The Arrhenius equation was computed as follows:

$$k = Ae^{\frac{-Ea}{RT}}$$

$$log(k) = log\left(Ae^{\frac{-Ea}{RT}}\right)$$

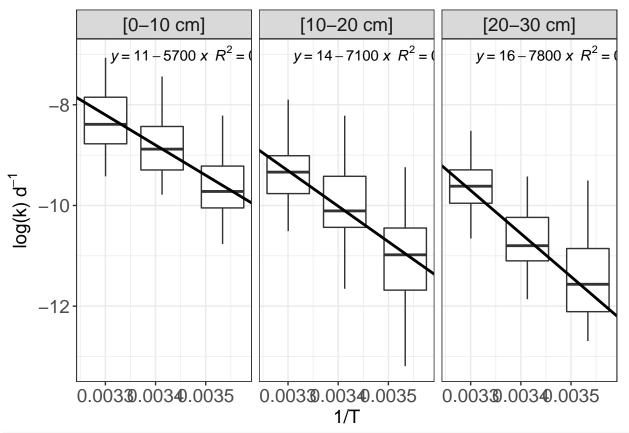
$$log(k) = log(A) + log\left(e^{\frac{-Ea}{RT}}\right)$$

$$log(k) = log(A) - \frac{1}{T} \times \left(\frac{Ea}{R}\right)$$

Where A is the pre-exponential factor and Ea is activation energy assumed to be independent of temperature, R is the universal gas constant and T is absolute temperature (Kelvin)

```
models_co2 <- data_co2 %>%
  group_by(`Depth (cm)`) %>%
  summarise(linmod = list(lm(lnK ~ `1/T`)))
models_co2
```

```
## # A tibble: 3 x 2
##
     `Depth (cm)` linmod
##
            <dbl> <list>
               10 <lm>
## 1
## 2
               20 <lm>
## 3
               30 < lm >
linmod_coef <- list()</pre>
for (i in seq_along(models_co2$linmod)) linmod_coef[[i]] <- models_co2$linmod[[i]]$coefficients
linmod_coef <- do.call(rbind.data.frame, linmod_coef)</pre>
names(linmod_coef) <- c("Intercept", "Slope")</pre>
linmod_coef <- bind_cols(unique(data_co2["Depth (cm)"]), linmod_coef)</pre>
linmod_coef
## # A tibble: 3 x 3
    `Depth (cm)` Intercept Slope
                      <dbl> <dbl>
##
            <dbl>
## 1
               10
                       11.6 -6002.
## 2
               20
                       14.0 -7052.
## 3
               30
                       18.5 -8558.
options(repr.plot.width = 12, repr.plot.height = 6)
plot_co2 <- data_co2_clean %>%
  ggplot(aes(x = 1/T), y = lnK)) +
  facet_grid(~`Depth (cm)`, labeller = labeller(`Depth (cm)` = New.labs_b)) +
  geom_boxplot(aes(group = factor(`1/T`)), outlier.shape = NA) +
  stat_regline_equation(aes(label = paste(..eq.label.., ..rr.label.., sep = "~~")), label.x = 0.00331,
  geom_abline(data = linmod_coef, aes(intercept = Intercept, slope = Slope), lwd = 1) +
  labs(y = bquote("log(k)" ~d^-1)) +
  theme_bw() +
  theme(strip.text = element_text(size = 14), axis.text=element_text(size=14),
        axis.title=element_text(size=14))
plot_co2
```



ggsave("figures/Arrhénius équation.png", plot = plot_co2, width = 8, height = 4, dpi = 600)# export plo

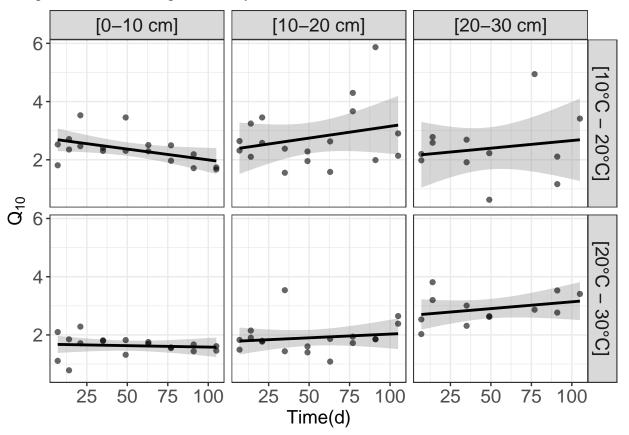
8.2 Activation Energy computation

```
Activation_energy <- tibble(</pre>
 Soil_layers = c("10", "20", "30"),
  intercept = NA,
  slope = NA,
  adj_r_sq = NA
)
lm_arrhenius <- for (i in 1:nrow(Activation_energy)) {</pre>
  lm_Activation_energy <- data_co2_clean %>%
    filter(`Depth (cm)` == Activation_energy$Soil_layers[i]) %>%
    lm(lnK \sim 1/T, data = .)
  # intercept
  Activation_energy$intercept[i] <- coef(lm_Activation_energy)[1]</pre>
  # Slope
  Activation_energy$slope[i] <- coef(lm_Activation_energy)[2]</pre>
  # statistics
  Activation_energy$adj_r_sq[i] <- summary(lm_Activation_energy)$adj.r.squared
}
```

```
mutate(Ea = -slope * R) %>%
  select(Soil_layers, adj_r_sq, Ea)
Activation_energy
## # A tibble: 3 x 3
    Soil_layers adj_r_sq
                   <dbl> <dbl>
##
     <chr>
                    0.477 47.5
## 1 10
## 2 20
                    0.402 58.6
## 3 30
                    0.507 64.9
8.3 Computing K median in order to compute Q10 value accross soil depth
K_median <- aggregate(K ~ `Sites` + `Time (days)` + `Depth (cm)` + `Temperature (°C)`, data = data_co2_</pre>
K_median <- K_median %>%
 pivot_wider(names_from = `Temperature (°C)`, values_from = K)
K_median$Q_20_10 <- K_median$`20` / K_median$`10`</pre>
K_median Q_30_20 \leftarrow K_median ^30 \ / K_median ^20
K_median <- K_median %>%
 na.omit(K_median)
data_Q10 <- gather(data = K_median, key = `Temperature range`, value = Q10, c(`Q_20_10`, `Q_30_20`),
                   factor_key=TRUE)
stat_Q10 <- data_Q10 |>
  group_by(`Depth (cm)`) |>
  get_summary_stats(Q10)
stat_Q10
## # A tibble: 3 x 14
    `Depth (cm)` variable
                             n min max median
                                                       q1
                                                             q3
                                                                  iqr
                                                                        mad mean
##
            <dbl> <chr>
                           <dbl> <
## 1
                              36 0.785 3.53
                                              1.82 1.67 2.32 0.647 0.542 1.99
               10 Q10
## 2
               20 Q10
                              36 1.08 5.87 2.05 1.79 2.63 0.845 0.579 2.33
                              24 0.627 4.94
                                               2.63 2.17 3.05 0.879 0.709 2.64
               30 Q10
## # ... with 3 more variables: sd <dbl>, se <dbl>, ci <dbl>
New.labs c <- c("[10°C - 20°C]", "[20°C - 30°C]") # Change labels
names(New.labs_c) \leftarrow c("Q_20_10", "Q_30_20")
options(repr.plot.width = 8, repr.plot.height = 4)
data Q10 |>
  mutate(`Layers` = as.character(`Depth (cm)`)) |>
  ggplot(aes(x = Time (days)), y = Q10)) +
  facet_grid(`Temperature range`~`Depth (cm)`,
             labeller = labeller(`Depth (cm)` = New.labs_b,
                                 `Temperature range` = New.labs_c)) +
  geom_smooth(method = "lm", se = TRUE, color = "Black") +
  geom_point(size = 1.5, alpha = 0.6) +
```

R = 8.3144621 / 1000 # Gas constant Kj/mol/K Activation_energy <- Activation_energy %>%

`geom_smooth()` using formula 'y ~ x'



ggsave("figures/Variation of Q10 across layers.png", width = 8, height = 4, dpi = 600)# export plot hig

`geom_smooth()` using formula 'y ~ x'

9. Soil description

9.1 Soil layers properties

Import data

```
data_carbon_credit <- data_carbon_credit |>
    mutate(`C:N ratio` = `Carbone (%)` / `Nitrogen (%)`)
Some calculations
mean_sd_CoverN <- data_carbon_credit |>
    group_by(`Layer (cm)`) |>
    summarize(mean_C_over_N = mean(`C:N ratio`, na.rm = TRUE),
                                               se_C_over_N = sd(`C:N ratio`, na.rm = TRUE)/sqrt(length(!is.na(`C:N ratio`))))
mean_sd_CoverN
## # A tibble: 3 x 3
##
          `Layer (cm)` mean_C_over_N se_C_over_N
##
          <chr>>
                                                   <dbl>
## 1 [0-10]
                                                   20.1
                                                                            1.05
                                                   16.0
## 2 [10-20]
                                                                            1.91
## 3 [20-30]
                                                     9.02
                                                                             1.96
data_carbon_credit |> get_summary_stats(`C stock (kg m-3)`)
## # A tibble: 1 x 13
         variable
                                                            max median
                                                                                        q1
                                                                                                    q3
                                                                                                              iqr
                                                                                                                       mad mean
                                                 \mathtt{min}
##
                                 <dbl> 
## 1 C stock (k~
                                       24 1.67 30.9
                                                                      12.5 6.57 16.7 10.1 7.78 12.1 7.05 1.44
## # ... with 1 more variable: ci <dbl>
data carbon credit |>
    group_by(`Layer (cm)`) |>
get_summary_stats(`C stock (kg m-3)`)
## # A tibble: 3 x 14
##
          `Layer (cm)` variable
                                                                                    max median
                                                                n
                                                                        \mathtt{min}
                                                                                                                q1
                                                                                                                            q3
                                                                                                                                      iqr
                                                                                                                                                 mad mean
##
          <chr>>
                                   <chr>
                                                         <dbl> <
## 1 [0-10]
                                   C stock (~
                                                                 8 11.4
                                                                                   22.2 16.2 15.1 17.4
                                                                                                                                    2.35 1.60 16.6
## 2 [10-20]
                                   C stock (~
                                                                 8 6.52 30.9 11.8
                                                                                                           6.74 17.1 10.4
                                                                                                                                                7.62 13.6
                                   C stock (~
                                                                8 1.67 14.8
                                                                                              5.36 3.31 7.42 4.11 3.04 6.09
## 3 [20-30]
## # ... with 3 more variables: sd <dbl>, se <dbl>, ci <dbl>
data_carbon_credit |>
    group_by(`Layer (cm)`) |>
get_summary_stats(`C:N ratio`)
## # A tibble: 3 x 14
          `Layer (cm)` variable
##
                                                              n
                                                                      min
                                                                                  max median
                                                                                                              q1
                                                                                                                          q3
                                                                                                                                    iqr
                                                                                                                                               mad mean
##
          <chr>>
                                   <chr>>
                                                       <dbl> <
## 1 [0-10]
                                   C:N ratio
                                                               8 15
                                                                                23.3 20.9 17.8
                                                                                                                      22.5 4.72 2.98 20.1
                                                                                 24.4 15.8 11.5
## 2 [10-20]
                                   C:N ratio
                                                               8 10
                                                                                                                      20
                                                                                                                                  8.5
                                                                                                                                              6.18 16.0
## 3 [20-30]
                                                                      2.5 20
                                                                                              8.33 5.96 10.6 4.66 4.32 9.02
                                   C:N ratio
                                                               8
## # ... with 3 more variables: sd <dbl>, se <dbl>, ci <dbl>
plot desc <- function(y, ylab){</pre>
   New.labs_c <- c("Site/A9", "Site/45") # Change labels</pre>
names(New.labs_c) <- c("Belanger/ A9", "Fortier/ 45")</pre>
    ggplot(data_carbon_credit, aes(`Layer (cm)`, y)) +
       geom boxplot() +
facet_grid( . ~ `Location`, scales = "free", labeller = labeller(`Location` = New.labs_c)) +
      theme_bw() +
```

```
theme(strip.text = element_text(size = 11), axis.text=element_text(size=11),
         axis.title=element text(size=11)) +
    labs(y = ylab)
  }
plot1 <- plot_desc(data_carbon_credit$`C stock (kg m-3)`, bquote("C stock (kg" ~m^-3~")"))</pre>
plot2 <- plot_desc(data_carbon_credit$`C:N ratio`, "C:N ratio")</pre>
plot3 <- plot desc(data carbon credit$`Bulk density (kg m-3)`, # m-3
                     "Bulk density (kg"~m^-3~")")
plot4 <- plot desc(data carbon credit$pHCaCl2, bquote(pHCaCl[2]))</pre>
plot5 <- plot_desc(data_carbon_credit$`Total porosity`, "Total porosity")</pre>
plot6 <- plot_desc(data_carbon_credit$`Water content (%)`, "Water content (%)")</pre>
options(repr.plot.width = 8, repr.plot.height = 6)
figure <- ggarrange(plot1, plot2, plot3, plot4, plot5, plot6,
                      labels = c("A", "B", "C", "D", "E", "F"), label.x = 0.05, label.y = 1.01,
                      ncol = 2, nrow = 3)
figure
C stock (kg m<sup>-3</sup>)
                                                    В
              Site/A9
                                  Site/45
                                                               Site/A9
                                                                                    Site/45
                                                  ratio
25
15
    30
    20
                                                  Z 10.
    10
        [0-10]10-2020-30] [0-10]10-2020-30]
                                                         [0-10]10-2020-30] [0-10]10-2020-30]
                      Layer (cm)
                                                                        Layer (cm)
Bulk density (kg m<sup>-3</sup>)
                                                    D
               Site/A9
                                                                Site/A9
                                   Site/45
                                                                                     Site/45
                                                  pHCaCl<sub>2</sub>
    1500
                                                     4.8
    1250
                                                     4.4
    1000
                                                     4.0
     750
          [0-10]0-2[22]0-30][0-10]0-2[22]0-30]
                                                          [0-10]10-2020-30] [0-10]10-2020-30]
                       Layer (cm)
                                                                        Layer (cm)
                                                  %F
  Ε
             Site/A9
                                  Site/45
                                                                Site/A9
                                                                                     Site/45
Total porosity
   0.7
                                                  Water content
                                                    0.50
                                                    0.45
   0.6
                                                     0.40
   0.5
                                                     0.35
  0.4
                                                     0.30
                                                           [0-10]10-2[2]0-30] [0-10]10-2[2]0-30]
       [0-10]10-2(240-30) [0-10]10-2(240-30]
                      Layer (cm)
                                                                         Layer (cm)
ggsave("figures/Soil description.png", width = 8, height = 5, dpi = 600) # export plot high resolution
```

9.2 C:N ratio in alternate sublayers of sand and organic matter

```
Data loading
```

```
Carbon_credit <- read_csv2('data/data_carbon_sublayer.csv')
```

```
## i Using "','" as decimal and "'.'" as grouping mark. Use `read_delim()` for more control.
## Rows: 23 Columns: 20
## -- Column specification ------
## Delimiter: ";"
## chr (6): Projet, Site, Horizon, Layers, Soil texture, Munsell_color
## dbl (14): Depht (cm), Thickness(cm), Bulk density(kg m-3), Site age, Weigh s...
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.
Carbon credit
## # A tibble: 23 x 20
                      Horizon `Depht (cm)` `Thickness(cm)` Layers `Bulk density(~`
     Projet Site
##
              <chr>
                                                    <dbl> <chr>
     <chr>
                                    <dbl>
                                                                            <dbl>
                      <chr>
                                                       1.8 [0-1.~
## 1 Pedology Belang~ H1
                                      1.8
                                                                             913.
## 2 Pedology Belang~ H2
                                       2.2
                                                      0.4 [1.8-~
                                                                             913.
## 3 Pedology Belang~ H3
                                       3.2
                                                      1 [2.2-~
                                                                             913.
## 4 Pedology Belang~ H4
                                       3.6
                                                      0.4 [3.2-~
                                                                             913.
## 5 Pedology Belang~ H5
                                      5.1
                                                      1.5 [3.6-~
                                                                             913.
## 6 Pedology Belang~ H6
                                      5.8
                                                      0.7 [5.1-~
                                                                             913.
## 7 Pedology Belang~ H7
                                      9.5
                                                      3.7 [5.8-~
                                                                             913.
## 8 Pedology Belang~ H8
                                      12
                                                      2 [9.5-~
                                                                            1384.
## 9 Pedology Belang~ H9
                                      12.5
                                                      0.5 [12-1~
                                                                            1384.
## 10 Pedology Belang~ H10
                                      19.2
                                                      6.7 [12.5~
                                                                            1384.
## # ... with 13 more rows, and 13 more variables: `Soil texture` <chr>,
      Site_age <dbl>, Munsell_color <chr>, Weigh_superior_2MM <dbl>,
      `Weigh _0_2MM` <dbl>, Repetition <dbl>, pHCaCl2 <dbl>, CTRL_C_pourc <dbl>,
      CTRL_S_pourc <dbl>, CTRL_N_pourc <dbl>, C_pourc <dbl>, S_pourc <dbl>,
## #
      N_pourc <dbl>
C:N ratio computation
Carbon_credit <- Carbon_credit %>%
mutate(`C/N` = C_pourc/N_pourc)
Generating the plots
options(repr.plot.width=8, repr.plot.height=4)
New.labs_d <- c("Site/A9", "Site/45") # Change labels</pre>
names(New.labs_d) <- c("Belanger/A9", "Fortier/45")</pre>
ggplot(data=Carbon_credit, aes(x= `Depht (cm)`, y= `C/N`)) +
 facet_grid(.~Site, labeller = labeller(`Site` = New.labs_d)) +
 geom_line(linetype = "twodash") +
 geom_point(aes(shape = `Soil texture`, fill = `Soil texture`), size = 3) +
 scale_shape_manual(values=c(21, 21))+
 scale_fill_manual(values = c("#000000", "#FFFFFF")) +
 scale_y_continuous(breaks = 5*0:1000,
                    expand = expand_scale(add = 5)) +
 scale_x_continuous(breaks = 5*0:1000,
```

theme(strip.text = element_text(face = "bold"), axis.text=element_text(face = "bold"),

axis.title=element_text(face = "bold") , legend.title= element_text(face = "bold"),

 $\frac{\text{expand}}{\text{expand}} = \frac{\text{expand}}{\text{scale}(\frac{\text{add}}{\text{expand}})} + \frac{\text{expand}}{\text{expand}}$

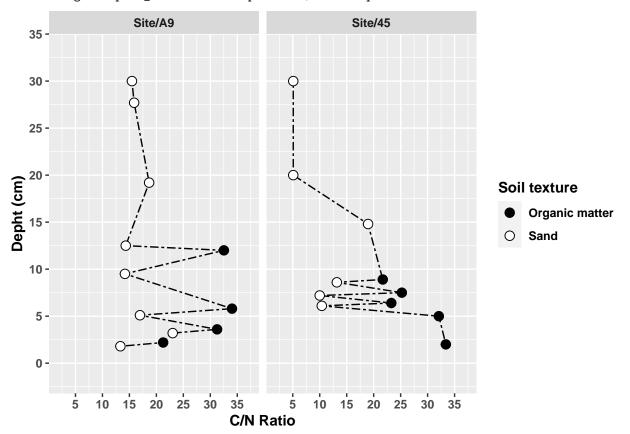
legend.text = element_text(face = "bold")) +

labs(y= "C/N Ratio") +

coord_flip()

```
## Warning: `expand_scale()` is deprecated; use `expansion()` instead.
```

Warning: `expand_scale()` is deprecated; use `expansion()` instead.



ggsave("figures/(C_over_N).png", width = 8, height = 4, dpi = 800)