mdi_pitchpine_analyses.R

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```
# script to analyze mdi pitch pine data

library(tidyverse)
library(emmeans)
library(lme4)
library(car)

multiplot <- function(..., plotlist=NULL, cols) {
  require(grid)

  # Make a list from the ... arguments and plotlist
  plots <- c(list(...), plotlist)

  numPlots = length(plots)

  # Make the panel
  plotCols = cols                          # Number of columns of plots
  plotRows = ceiling(numPlots/plotCols) # Number of rows needed, calculated from # of cols

  # Set up the page
  grid.newpage()
  pushViewport(viewport(layout = grid.layout(plotRows, plotCols)))
  vplayout <- function(x, y)
    viewport(layout.pos.row = x, layout.pos.col = y)

  # Make each plot, in the correct location
  for (i in 1:numPlots) {
    curRow = ceiling(i/plotCols)
    curCol = (i-1) %% plotCols + 1
    print(plots[[i]], vp = vplayout(curRow, curCol ))
  }

}

## read in cleaned data
data = read.csv('../data/mdi_all_clean.csv')
data$CN_foliar = data$C_foliar/data$N_foliar
data$CN_soil = data$C_soil/data$N_soil
data$fire[data$Name == 'CAD'] = 'fire'
data$fire[data$Name == 'CADCLIFFS'] = 'fire'
data$fire[data$Name == 'STSAUV'] = 'no fire'
data$fire[data$Name == 'WOND'] = 'no fire'
head(data)
```

```
##   X                ID Name height canopy diam   d13C  d15N C_foliar
## 1 1 PP-1-LOWELEV-DIST WOND  472.4  548.6 21.6 -28.52 -5.98    47.75
## 2 2 PP-2-LOWELEV-DIST WOND  152.4  167.6 19.4 -28.89 -1.78    48.58
```

```
## 3 3 PP-3-LOWELEV-DIST WOND  365.8  365.8 20.3 -29.14 -6.81    50.39
## 4 4 PP-4-LOWELEV-DIST WOND  365.8  609.6 20.3 -27.65 -6.81    47.37
## 5 5 PP-5-LOWELEV-DIST WOND  487.7  557.6 20.3 -28.19 -1.33    36.73
## 6 6 PP-6-LOWELEV-DIST WOND  762.0  731.5 21.6 -29.58  1.24    25.92
##   N_foliar Ca_foliar P_foliar K_foliar Mg_foliar Al_foliar Zn_foliar
## 1     1.07      1860      830     3720       910       320      51.5
## 2     1.79      1420     2260      540       330       176      27.7
## 3    18.37      1710      930     3430       880       606      34.0
## 4     1.05      1220     2700      460       570       274      24.7
## 5     0.67      1480      810     3170       890       488      29.7
## 6     1.02      1230     2020      630       440       168      25.9
##   Ca_soil P_soil K_soil Mg_soil Al_soil Zn_soil  pH  CEC C_soil N_soil
## 1    1506   19.1    627     635     270    10.9 3.9 34.3  47.27   0.95
## 2    1007   12.4    394     416      38     8.6 3.5 33.1  32.15   0.91
## 3     408    1.4     79      71     124     1.9 4.5 15.9  27.17   0.64
## 4     476    1.5    183     207      86     3.2 4.0 18.7  18.62   0.31
## 5     926   13.8    475     327      92     6.4 3.6 33.9  12.70   0.34
## 6     764    9.2    420     290      77     0.4 3.8 23.6  23.36   0.12
##    ID1 longitude latitude Label Elevation Slope Aspect Compass retention
## 1 PP-1 -68.31526 44.23166   PP1        17     4    287    West      25.4
## 2 PP-2 -68.31491 44.23153   PP2        18     0    323      NE      14.8
## 3 PP-3 -68.31337 44.23086   PP3        12     2    120      SE      19.1
## 4 PP-4 -68.31492 44.23148   PP4        18     2    205      SW       9.3
## 5 PP-5 -68.31441 44.23136   PP5        17     5    111    East      18.6
## 6 PP-6 -68.31442 44.23117   PP6        15     3     56      NE      31.9
##   CN_foliar   CN_soil    fire
## 1 44.626168  49.75789 no fire
## 2 27.139665  35.32967 no fire
## 3  2.743059  42.45312 no fire
## 4 45.114286  60.06452 no fire
## 5 54.820896  37.35294 no fire
## 6 25.411765 194.66667 no fire
```

```r
## site means
data_group_by_Name = group_by(data, Name)
data_Name_means = summarise(data_group_by_Name,
                            Elevation_mean = mean(Elevation, na.rm = T),
                            Slope_mean = mean(Slope, na.rm = T),
                            Aspect_mean = mean(Aspect, na.rm = T))

## create an elevation factor
data$elevation_fac[data$Name == 'CAD' | data$Name == 'STSAUV'] = 'high'
data$elevation_fac[data$Name == 'CADCLIFFS' | data$Name == 'WOND'] = 'low'

## create a generic variable set to pass to formula argument
ind_variables = c('elevation_fac', 'fire')
dep_variables = c("log(Elevation)", "log(height)", "log(canopy)", "log(diam)",
                  "d13C", "d15N", "C_foliar", "N_foliar", "CN_foliar", "Ca_foliar", "log(P_foliar)",
                  "log(K_foliar)", "Mg_foliar", "Al_foliar", "log(Zn_foliar)",
                  "Ca_soil", "log(P_soil)", "K_soil", "Mg_soil", "log(Al_soil)", "log(Zn_soil)",
                  "pH", "CEC", "C_soil", "N_soil", "log(CN_soil)", "asin(sqrt(0.01 * retention))")

## fit models and explore results
```

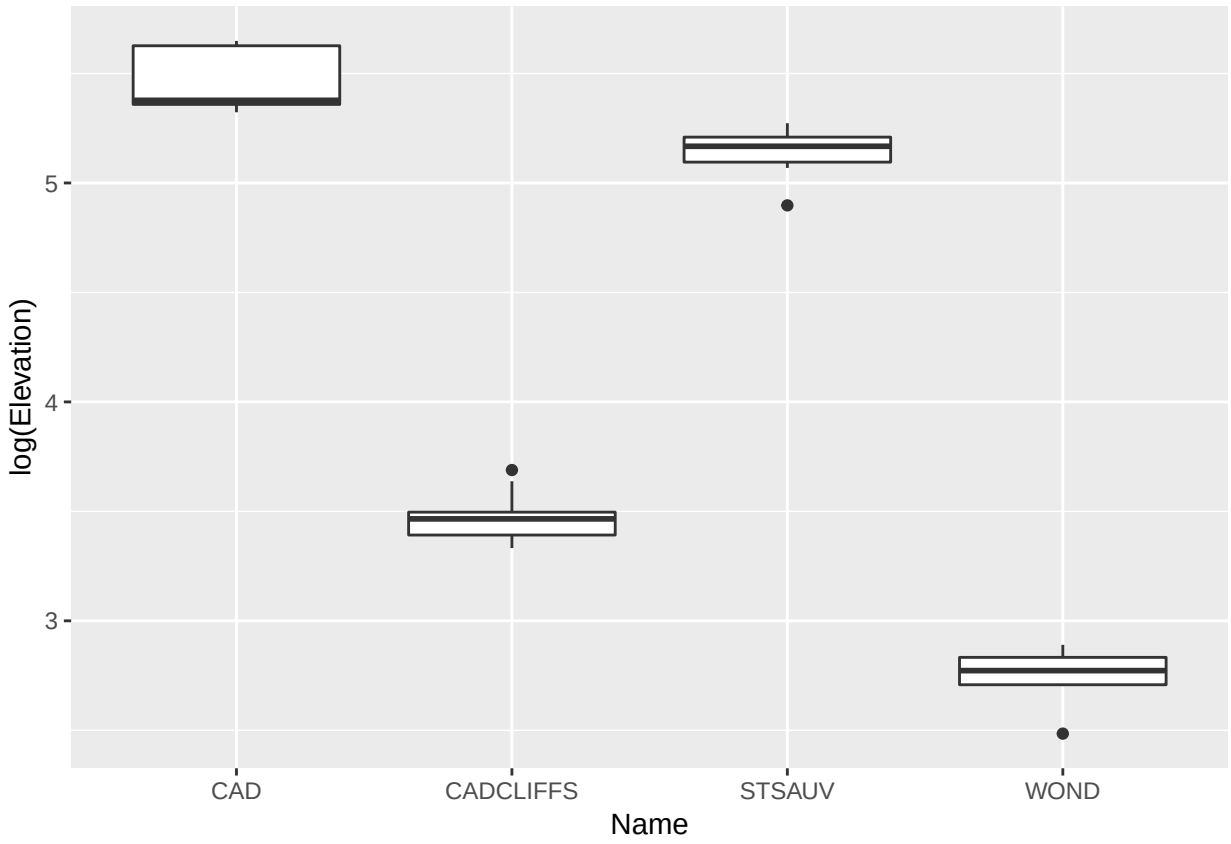
```r
### elevation
Elevation_lm = lm(as.formula(paste(dep_variables[1],
                                   paste(ind_variables, collapse = "*"),
                                   sep = "~")), data = data)
#plot(resid(Elevation_lm) ~ fitted(Elevation_lm))
Anova(Elevation_lm)
```

```
## Anova Table (Type II tests)
##
## Response: log(Elevation)
##                     Sum Sq Df  F value    Pr(>F)
## elevation_fac      47.734  1 3241.737 < 2.2e-16 ***
## fire                2.740  1  186.101 8.561e-16 ***
## elevation_fac:fire  0.379  1   25.751 1.196e-05 ***
## Residuals           0.530 36
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```r
cld(emmeans(Elevation_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean     SE df lower.CL upper.CL .group
##  low           no fire   2.76 0.0384 36     2.68     2.84  1
##  low           fire      3.48 0.0384 36     3.40     3.56   2
##  high          no fire   5.14 0.0384 36     5.06     5.22    3
##  high          fire      5.47 0.0384 36     5.39     5.55     4
##
## Results are given on the log (not the response) scale.
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```r
ggplot(data = data, aes(x = Name, y = log(Elevation))) +
  geom_boxplot()
```




```
### height
height_lm = lm(as.formula(paste(dep_variables[2],
                                paste(ind_variables, collapse = "*"),
                                sep = "~")), data = data)
#plot(resid(height_lm) ~ fitted(height_lm))
Anova(height_lm)
```

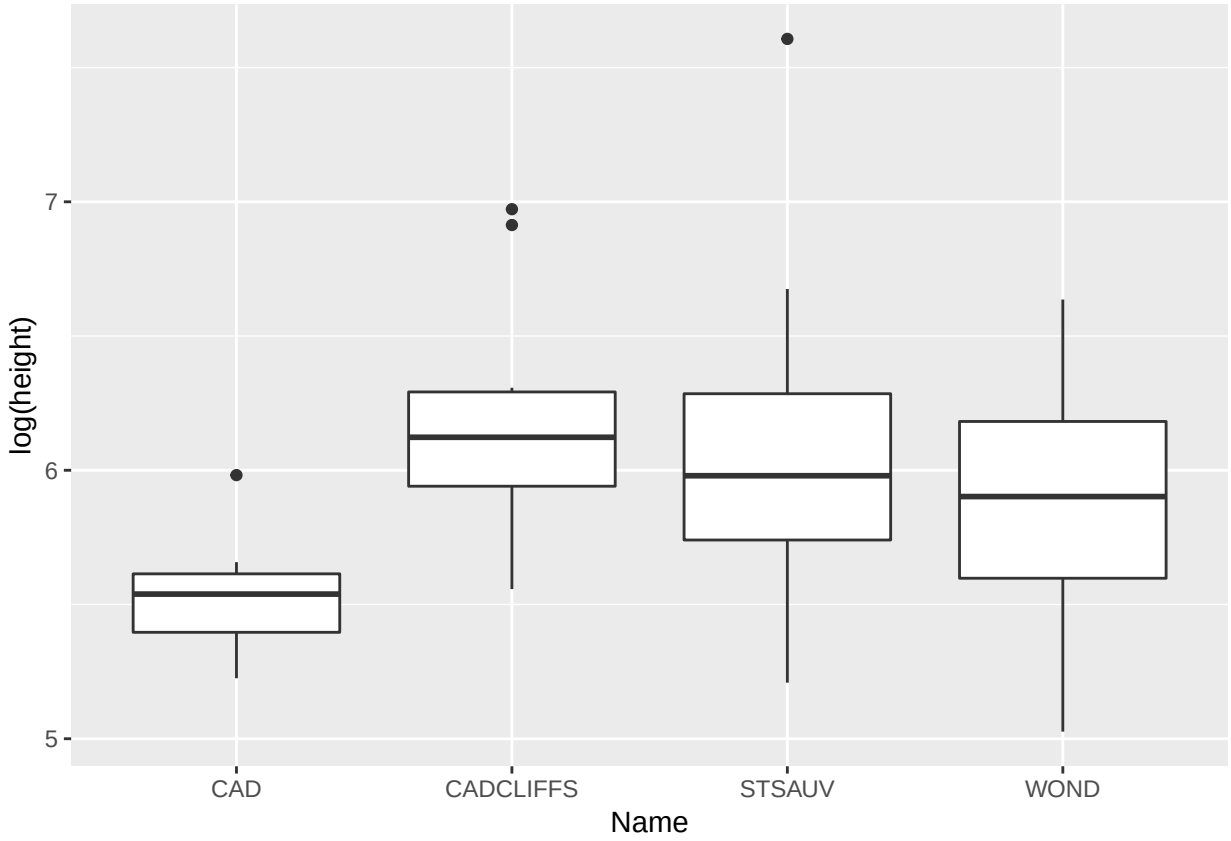
```
## Anova Table (Type II tests)
##
## Response: log(height)
##                     Sum Sq Df F value   Pr(>F)
## elevation_fac      0.5238  1  2.2201 0.144934
## fire               0.1207  1  0.5116 0.479068
## elevation_fac:fire 1.9021  1  8.0628 0.007385 **
## Residuals          8.4928 36
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
cld(emmeans(height_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean    SE df lower.CL upper.CL .group
##  high          fire      5.54 0.154 36     5.22     5.85  1
##  low           no fire   5.88 0.154 36     5.56     6.19  12
##  high          no fire   6.08 0.154 36     5.77     6.39  12
##  low           fire      6.20 0.154 36     5.89     6.51   2
##
## Results are given on the log (not the response) scale.
## Confidence level used: 0.95
```

```
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = log(height))) +
  geom_boxplot()
```




```
# height_lmer_cont = lmer(log(height) ~ Elevation * fire + (1|Name), data = data)
# Anova(height_lmer_cont)
# test(emtrends(height_lmer_cont, ~fire, var = 'Elevation'))

# height_plot = ggplot(data = data, aes(x = Name, y = log(height), col = fire)) +
#   theme(legend.position = "none",
#         axis.title.y=element_text(size=rel(2.5), colour = 'black'),
#         axis.title.x=element_text(size=rel(2.5), colour = 'black'),
#         axis.text.x=element_text(size=rel(2), colour = 'black'),
#         axis.text.y=element_text(size=rel(2), colour = 'black'),
#         panel.background = element_rect(fill = 'white', colour = 'black'),
#         panel.grid.major = element_line(colour = "grey")) +
#   geom_boxplot(outlier.color = NA, fill = 'white') +
#   geom_dotplot(binaxis = 'y', binwidth = 0.07, stackdir = 'center', alpha = 0.5) +
#   # scale_x_discrete(labels = c('Ambient', 'Added N')) +
#   xlab('Site') +
#   ylab(expression('ln(Height)'))
#
# height_plot_elevation = ggplot(data = data, aes(x = Elevation, y = log(height), col = fire)) +
#   theme(legend.position = "right",
#         axis.title.y=element_text(size=rel(2.5), colour = 'black'),
#         axis.title.x=element_text(size=rel(2.5), colour = 'black'),
```

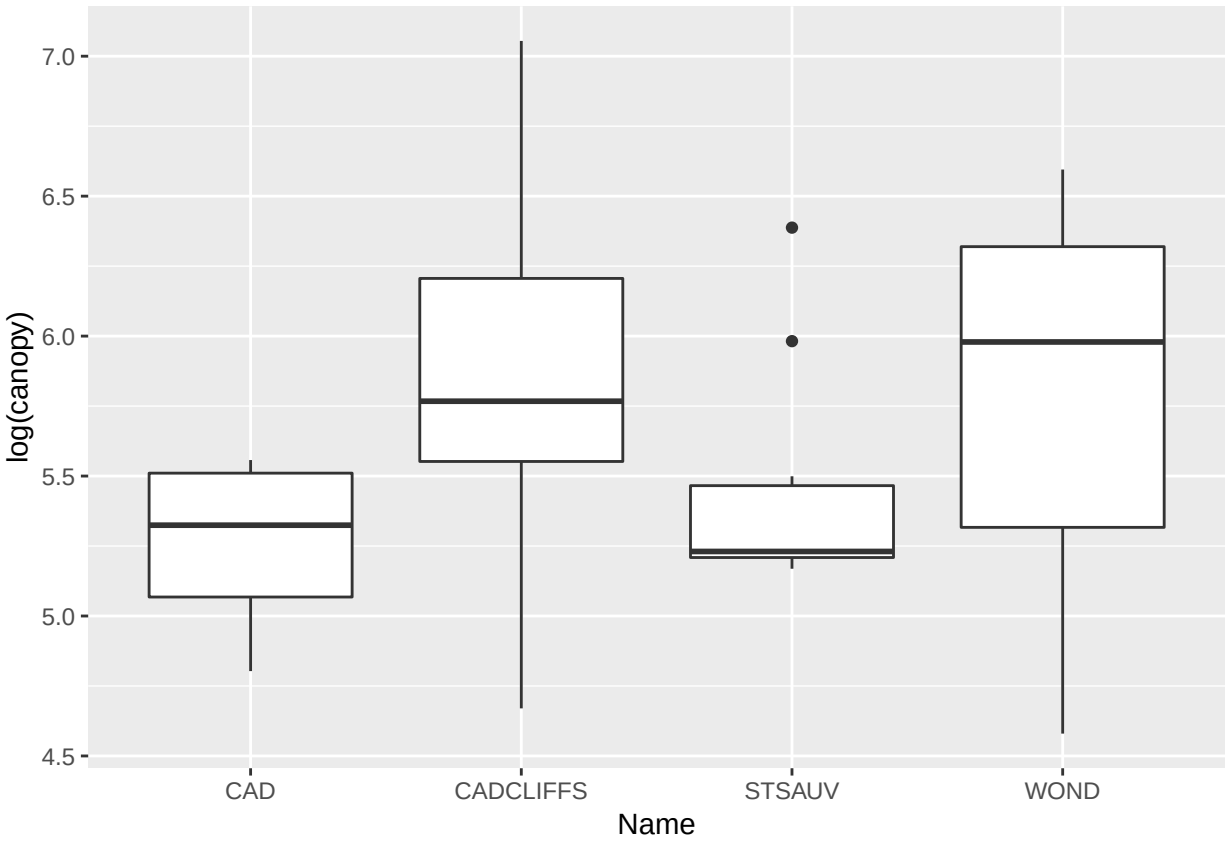
```
#         axis.text.x=element_text(size=rel(2), colour = 'black'),
#         axis.text.y=element_text(size=rel(2), colour = 'black'),
#         panel.background = element_rect(fill = 'white', colour = 'black'),
#         panel.grid.major = element_line(colour = "grey")) +
#   geom_point(size = 6) +
#   ylab(expression('ln(Height)'))
#
# jpeg(filename = "plots/height_plot.jpeg", width = 1000, height = 600, units = 'px')
# multiplot(height_plot, height_plot_elevation, cols = 2)
# dev.off()

### canopy
canopy_lm = lm(as.formula(paste(dep_variables[3],
                                paste(ind_variables, collapse = "*"),
                                sep = "~")), data = data)
#plot(resid(canopy_lm) ~ fitted(canopy_lm))
anova(canopy_lm)
```

```
## Analysis of Variance Table
##
## Response: log(canopy)
##                     Df  Sum Sq Mean Sq F value   Pr(>F)
## elevation_fac        1  2.3211 2.32110  8.3352 0.006538 **
## fire                 1  0.0673 0.06729  0.2416 0.626008
## elevation_fac:fire   1  0.1170 0.11701  0.4202 0.520961
## Residuals           36 10.0249 0.27847
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
cld(emmeans(canopy_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean    SE df lower.CL upper.CL .group
##  high          fire      5.26 0.167 36     4.92     5.60  1
##  high          no fire   5.45 0.167 36     5.11     5.79  1
##  low           no fire   5.82 0.167 36     5.48     6.16  1
##  low           fire      5.85 0.167 36     5.51     6.19  1
##
## Results are given on the log (not the response) scale.
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = log(canopy))) +
  geom_boxplot()
```




```
# canopy_lmer_cont = lmer(log(canopy) ~ Elevation * fire + (1|Name), data = data)
# Anova(canopy_lmer_cont)
# test(emtrends(canopy_lmer_cont, ~fire, var = 'Elevation'))

# canopy_plot = ggplot(data = data, aes(x = Name, y = log(canopy), col = fire)) +
#   theme(legend.position = "none",
#         axis.title.y=element_text(size=rel(2.5), colour = 'black'),
#         axis.title.x=element_text(size=rel(2.5), colour = 'black'),
#         axis.text.x=element_text(size=rel(2), colour = 'black'),
#         axis.text.y=element_text(size=rel(2), colour = 'black'),
#         panel.background = element_rect(fill = 'white', colour = 'black'),
#         panel.grid.major = element_line(colour = "grey")) +
#   geom_boxplot(outlier.color = NA, fill = 'white') +
#   geom_dotplot(binaxis = 'y', binwidth = 0.07, stackdir = 'center', alpha = 0.5) +
#   # scale_x_discrete(labels = c('Ambient', 'Added N')) +
#   xlab('Site') +
#   ylab(expression('ln(Canopy)'))
#
# canopy_plot_elevation = ggplot(data = data, aes(x = Elevation, y = log(canopy), col = fire)) +
#   theme(legend.position = "right",
#         axis.title.y=element_text(size=rel(2.5), colour = 'black'),
#         axis.title.x=element_text(size=rel(2.5), colour = 'black'),
#         axis.text.x=element_text(size=rel(2), colour = 'black'),
#         axis.text.y=element_text(size=rel(2), colour = 'black'),
#         panel.background = element_rect(fill = 'white', colour = 'black'),
#         panel.grid.major = element_line(colour = "grey")) +
#   geom_point(size = 6) +
```

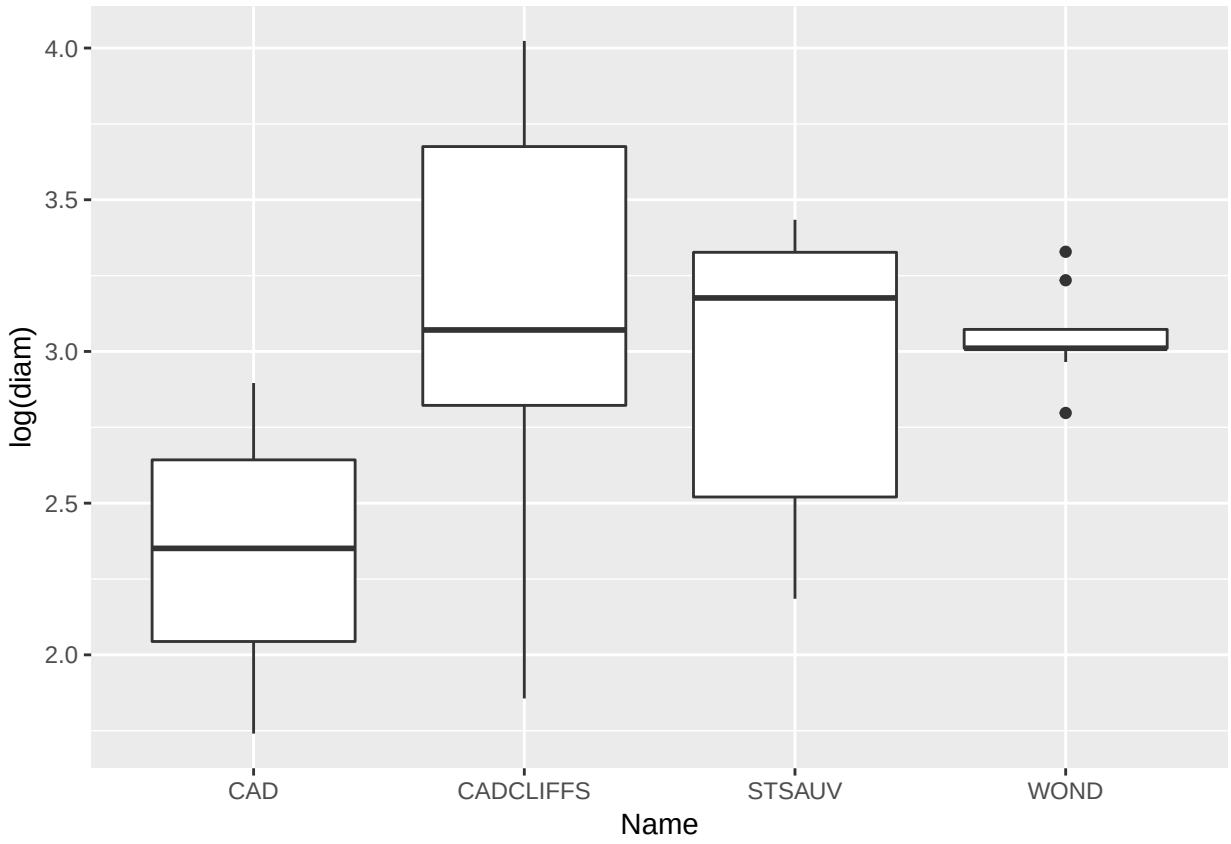
```
#   ylab(expression('ln(Canopy)'))
#
# jpeg(filename = "plots/canopy_plot.jpeg", width = 1000, height = 600, units = 'px')
# multiplot(canopy_plot, canopy_plot_elevation, cols = 2)
# dev.off()

### diam
diam_lm = lm(as.formula(paste(dep_variables[4],
                              paste(ind_variables, collapse = "*"),
                              sep = "~")), data = data)
#plot(resid(diam_lm) ~ fitted(diam_lm))
anova(diam_lm)
```

```
## Analysis of Variance Table
##
## Response: log(diam)
##                    Df Sum Sq Mean Sq F value   Pr(>F)
## elevation_fac       1 2.1167 2.11673 10.0838 0.003062 **
## fire                1 0.7177 0.71769  3.4190 0.072675 .
## elevation_fac:fire  1 1.2783 1.27835  6.0899 0.018483 *
## Residuals          36 7.5569 0.20991
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
cld(emmeans(diam_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean    SE df lower.CL upper.CL .group
##  high          fire      2.32 0.145 36     2.03     2.62  1
##  high          no fire   2.95 0.145 36     2.66     3.24   2
##  low           no fire   3.05 0.145 36     2.76     3.35   2
##  low           fire      3.14 0.145 36     2.85     3.43   2
##
## Results are given on the log (not the response) scale.
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = log(diam))) +
  geom_boxplot()
```




```
# diam_lmer_cont = lmer(log(diam) ~ Elevation * fire + (1|Name), data = data)
# Anova(diam_lmer_cont)
# test(emtrends(diam_lmer_cont, ~fire, var = 'Elevation'))

# diam_plot = ggplot(data = data, aes(x = Name, y = log(diam), col = fire)) +
#   theme(legend.position = "none",
#         axis.title.y=element_text(size=rel(2.5), colour = 'black'),
#         axis.title.x=element_text(size=rel(2.5), colour = 'black'),
#         axis.text.x=element_text(size=rel(2), colour = 'black'),
#         axis.text.y=element_text(size=rel(2), colour = 'black'),
#         panel.background = element_rect(fill = 'white', colour = 'black'),
#         panel.grid.major = element_line(colour = "grey")) +
#   geom_boxplot(outlier.color = NA, fill = 'white') +
#   geom_dotplot(binaxis = 'y', binwidth = 0.07, stackdir = 'center', alpha = 0.5) +
#   # scale_x_discrete(labels = c('Ambient', 'Added N')) +
#   xlab('Site') +
#   ylab(expression('ln(Diameter)'))
#
# diam_plot_elevation = ggplot(data = data, aes(x = Elevation, y = log(diam), col = fire)) +
#   theme(legend.position = "right",
#         axis.title.y=element_text(size=rel(2.5), colour = 'black'),
#         axis.title.x=element_text(size=rel(2.5), colour = 'black'),
#         axis.text.x=element_text(size=rel(2), colour = 'black'),
#         axis.text.y=element_text(size=rel(2), colour = 'black'),
#         panel.background = element_rect(fill = 'white', colour = 'black'),
#         panel.grid.major = element_line(colour = "grey")) +
#   geom_point(size = 6) +
```

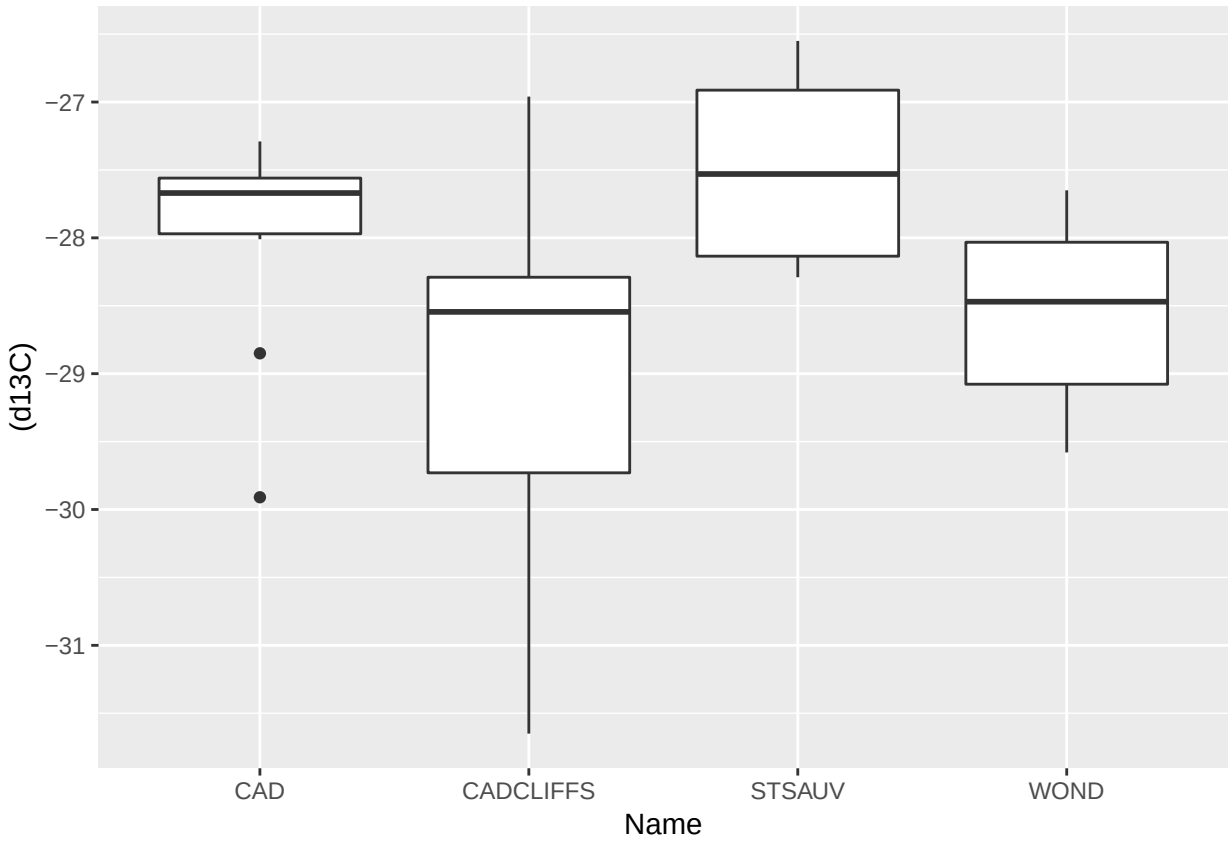
```
#   ylab(expression('ln(Diameter)'))
#
# jpeg(filename = "plots/diam_plot.jpeg", width = 1000, height = 600, units = 'px')
# multiplot(diam_plot, diam_plot_elevation, cols = 2)
# dev.off()

### d13C
d13C_lm = lm(as.formula(paste(dep_variables[5],
                              paste(ind_variables, collapse = "*"),
                              sep = "~")), data = data)
#plot(resid(d13C_lm) ~ fitted(d13C_lm))
anova(d13C_lm)
```

```
## Analysis of Variance Table
##
## Response: d13C
##                    Df  Sum Sq Mean Sq F value   Pr(>F)
## elevation_fac       1 10.6193 10.6193 12.2725 0.001248 **
## fire                1  2.2420  2.2420  2.5911 0.116204
## elevation_fac:fire  1  0.0006  0.0006  0.0007 0.979800
## Residuals          36 31.1505  0.8653
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
cld(emmeans(d13C_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean    SE df lower.CL upper.CL .group
##  low           fire     -29.0 0.294 36    -29.6    -28.4  1
##  low           no fire  -28.5 0.294 36    -29.1    -27.9  12
##  high          fire     -28.0 0.294 36    -28.6    -27.4  12
##  high          no fire  -27.5 0.294 36    -28.1    -26.9   2
##
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (d13C))) +
  geom_boxplot()
```



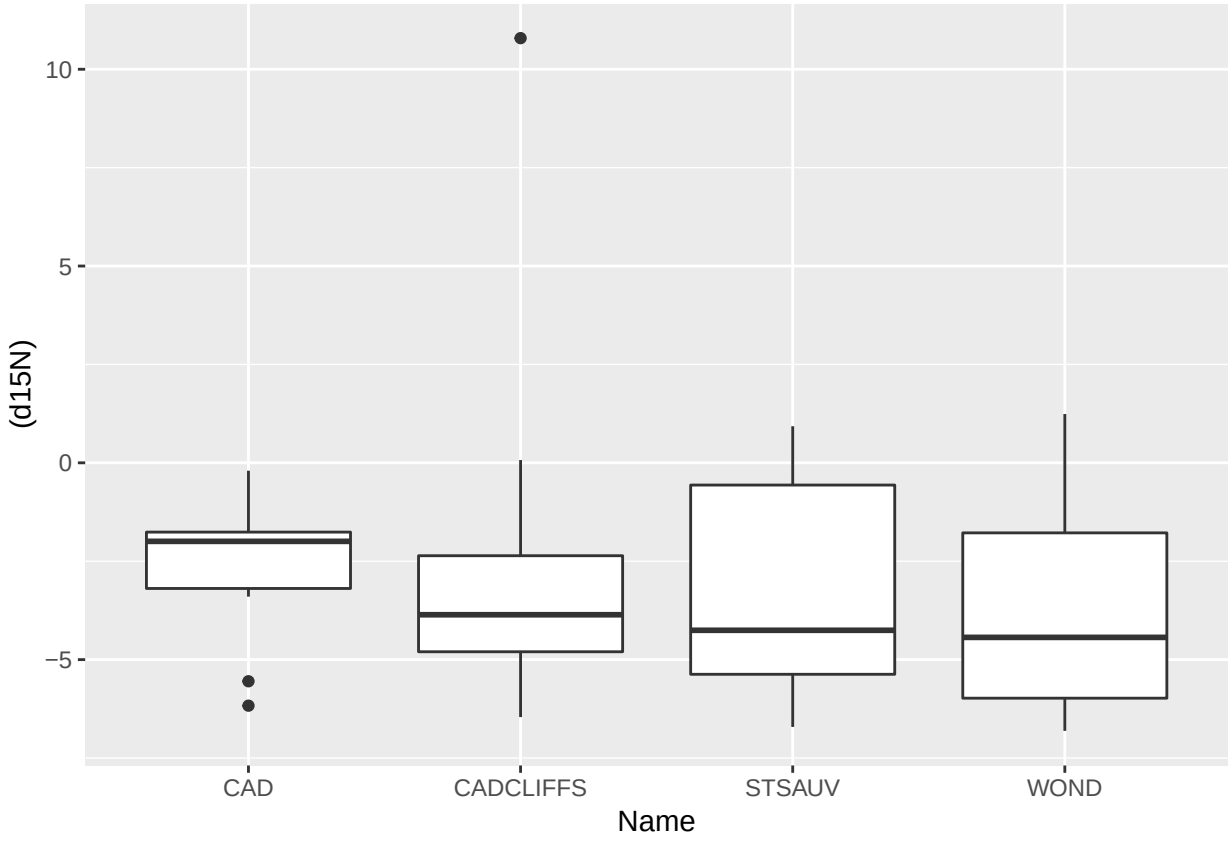

```
### d15N
d15N_lm = lm(as.formula(paste(dep_variables[6],
                              paste(ind_variables, collapse = "*"),
                              sep = "~")), data = data)
#plot(resid(d15N_lm) ~ fitted(d15N_lm))
anova(d15N_lm)
```

```
## Analysis of Variance Table
##
## Response: d15N
##                    Df Sum Sq Mean Sq F value Pr(>F)
## elevation_fac       1   0.31  0.3098  0.0273 0.8697
## fire                1  11.32 11.3210  0.9973 0.3246
## elevation_fac:fire  1   1.13  1.1290  0.0994 0.7543
## Residuals          36 408.68 11.3522
```

```
cld(emmeans(d15N_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean   SE df lower.CL upper.CL .group
##  low           no fire  -3.81 1.07 36    -5.97   -1.649  1
##  high          no fire  -3.30 1.07 36    -5.46   -1.137  1
##  high          fire     -2.57 1.07 36    -4.73   -0.409  1
##  low           fire     -2.41 1.07 36    -4.57   -0.249  1
##
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (d15N))) +
  geom_boxplot()
```




```
### C_foliar
C_foliar_lm = lm(as.formula(paste(dep_variables[7],
                                  paste(ind_variables, collapse = "*"),
                                  sep = "~")), data = data)
#plot(resid(C_foliar_lm) ~ fitted(C_foliar_lm))
anova(C_foliar_lm)
```

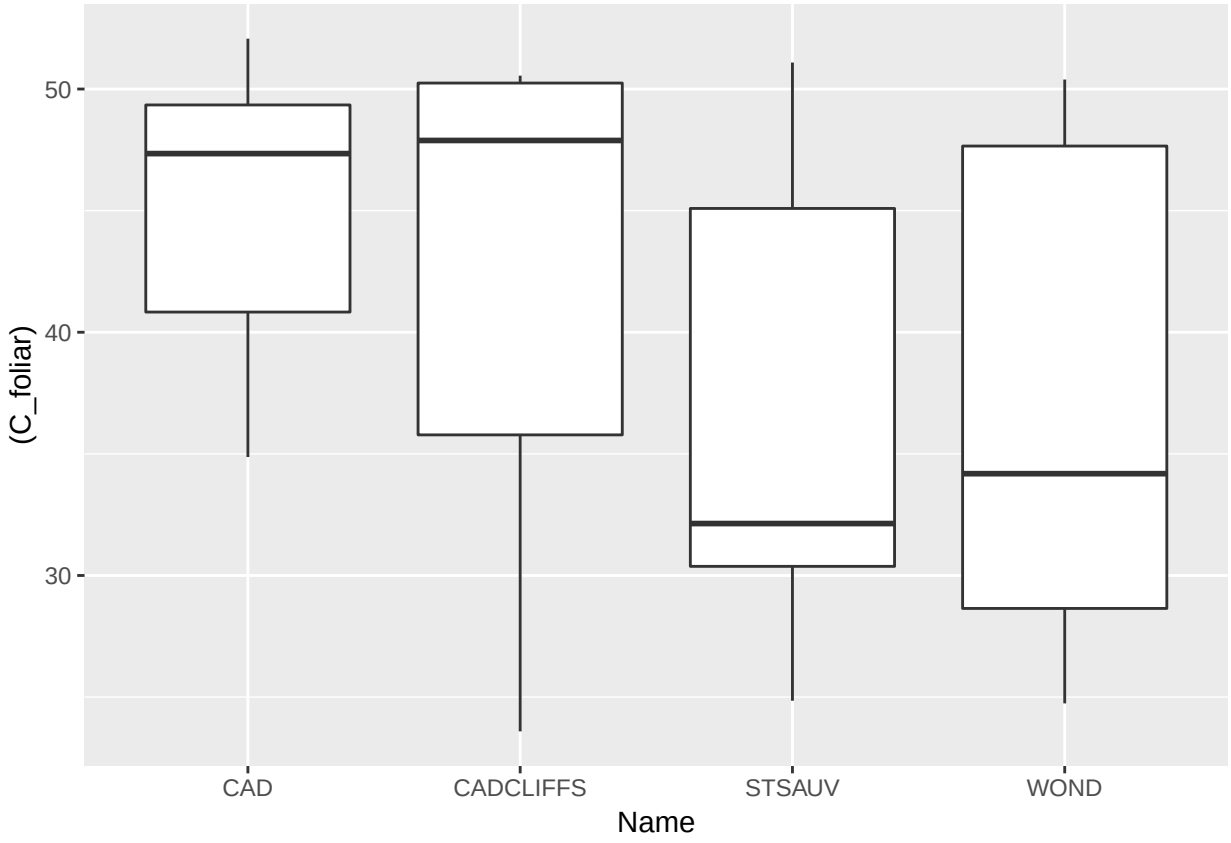
```
## Analysis of Variance Table
##
## Response: C_foliar
##                     Df  Sum Sq Mean Sq F value  Pr(>F)
## elevation_fac        1    2.87    2.87  0.0353 0.85201
## fire                 1  553.91  553.91  6.8202 0.01306 *
## elevation_fac:fire   1   25.46   25.46  0.3134 0.57904
## Residuals           36 2923.77   81.22
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
cld(emmeans(C_foliar_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean   SE df lower.CL upper.CL .group
##  high          no fire   36.0 2.85 36     30.2     41.8  1
##  low           no fire   37.1 2.85 36     31.3     42.9  1
##  low           fire      42.9 2.85 36     37.2     48.7  1
##  high          fire      45.1 2.85 36     39.3     50.8  1
```

```
##
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (C_foliar))) +
  geom_boxplot()
```




```
### N_foliar
N_foliar_lm = lm(as.formula(paste(dep_variables[8],
                                  paste(ind_variables, collapse = "*"),
                                  sep = "~")), data = data)
#plot(resid(N_foliar_lm) ~ fitted(N_foliar_lm))
anova(N_foliar_lm)
```

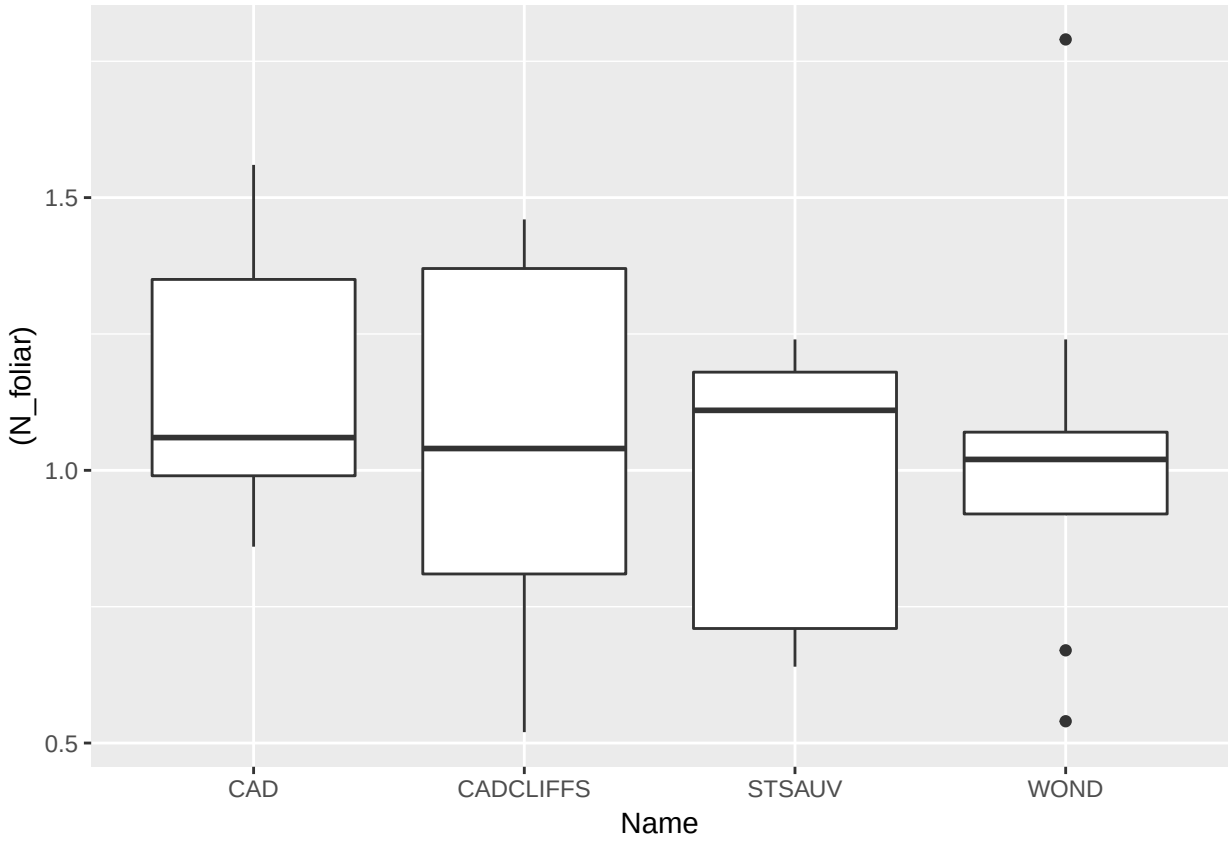
```
## Analysis of Variance Table
##
## Response: N_foliar
##                     Df  Sum Sq Mean Sq F value Pr(>F)
## elevation_fac        1    1.06   1.056  0.0292 0.8652
## fire                 1    0.38   0.376  0.0104 0.9193
## elevation_fac:fire   1    0.27   0.269  0.0074 0.9317
## Residuals           36 1300.52  36.125
```

```
cld(emmeans(N_foliar_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean  SE df lower.CL upper.CL .group
##  low           no fire   2.76 1.9 36   -1.095     6.61  1
##  low           fire      2.79 1.9 36   -1.065     6.64  1
```

```
##  high          no fire   2.92 1.9 36   -0.934     6.78  1
##  high          fire      3.28 1.9 36   -0.576     7.13  1
##
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = subset(data, N_foliar < 5), aes(x = Name, y = (N_foliar))) +
  geom_boxplot()
```




```
### CN_foliar
CN_foliar_lm = lm(as.formula(paste(dep_variables[9],
                                   paste(ind_variables, collapse = "*"),
                                   sep = "~")), data = data)
#plot(resid(CN_foliar_lm) ~ fitted(CN_foliar_lm))
anova(CN_foliar_lm)
```

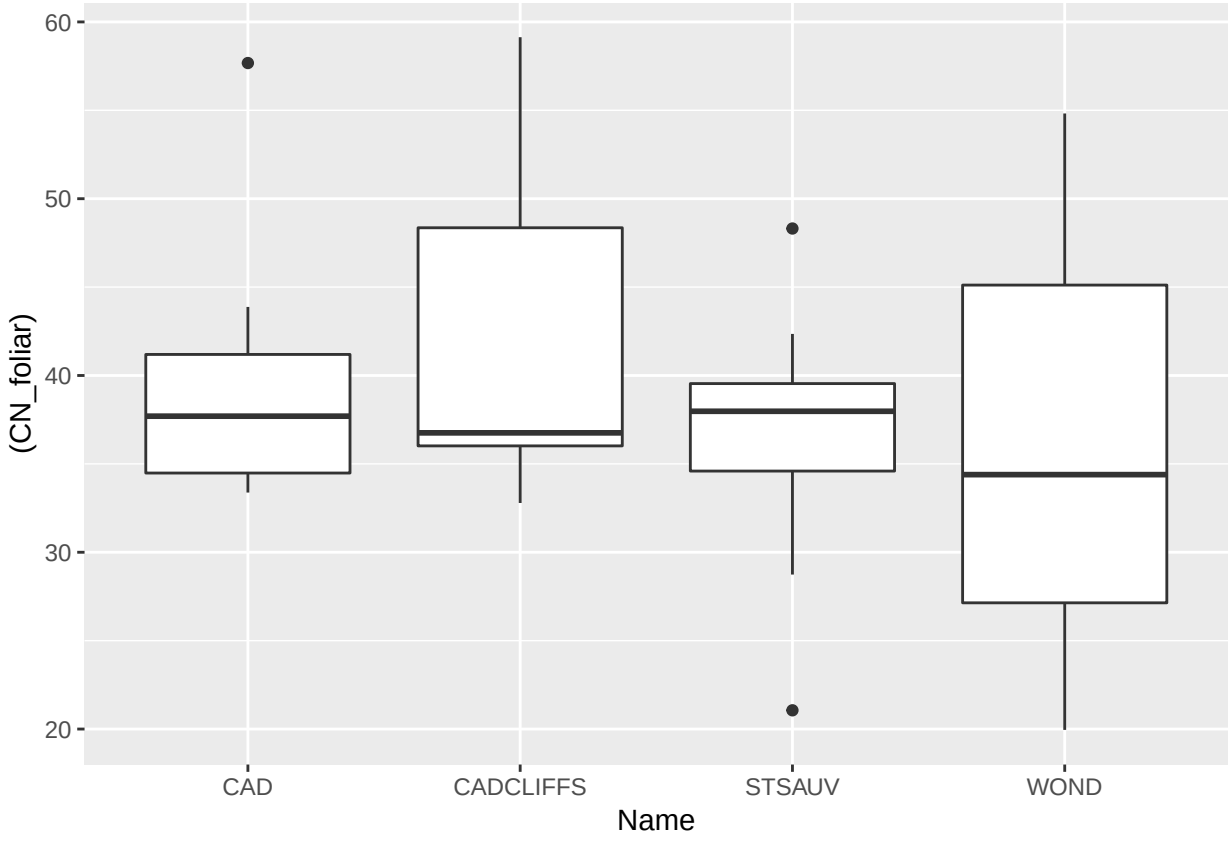
```
## Analysis of Variance Table
##
## Response: CN_foliar
##                    Df Sum Sq Mean Sq F value Pr(>F)
## elevation_fac       1   33.5  33.477  0.1542 0.6969
## fire                1  143.2 143.250  0.6599 0.4219
## elevation_fac:fire  1   12.5  12.493  0.0575 0.8118
## Residuals          36 7815.1 217.087
```

```
cld(emmeans(CN_foliar_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean   SE df lower.CL upper.CL .group
```

```
##  high          no fire   33.2 4.66 36     23.7     42.6  1
##  low           no fire   33.9 4.66 36     24.5     43.4  1
##  high          fire      35.9 4.66 36     26.4     45.3  1
##  low           fire      38.8 4.66 36     29.4     48.3  1
##
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = subset(data, N_foliar < 5), aes(x = Name, y = (CN_foliar))) +
  geom_boxplot()
```




```
### Ca_foliar
Ca_foliar_lm = lm(as.formula(paste(dep_variables[10],
                                   paste(ind_variables, collapse = "*"),
                                   sep = "~")), data = data)
#plot(resid(Ca_foliar_lm) ~ fitted(Ca_foliar_lm))
anova(Ca_foliar_lm)
```

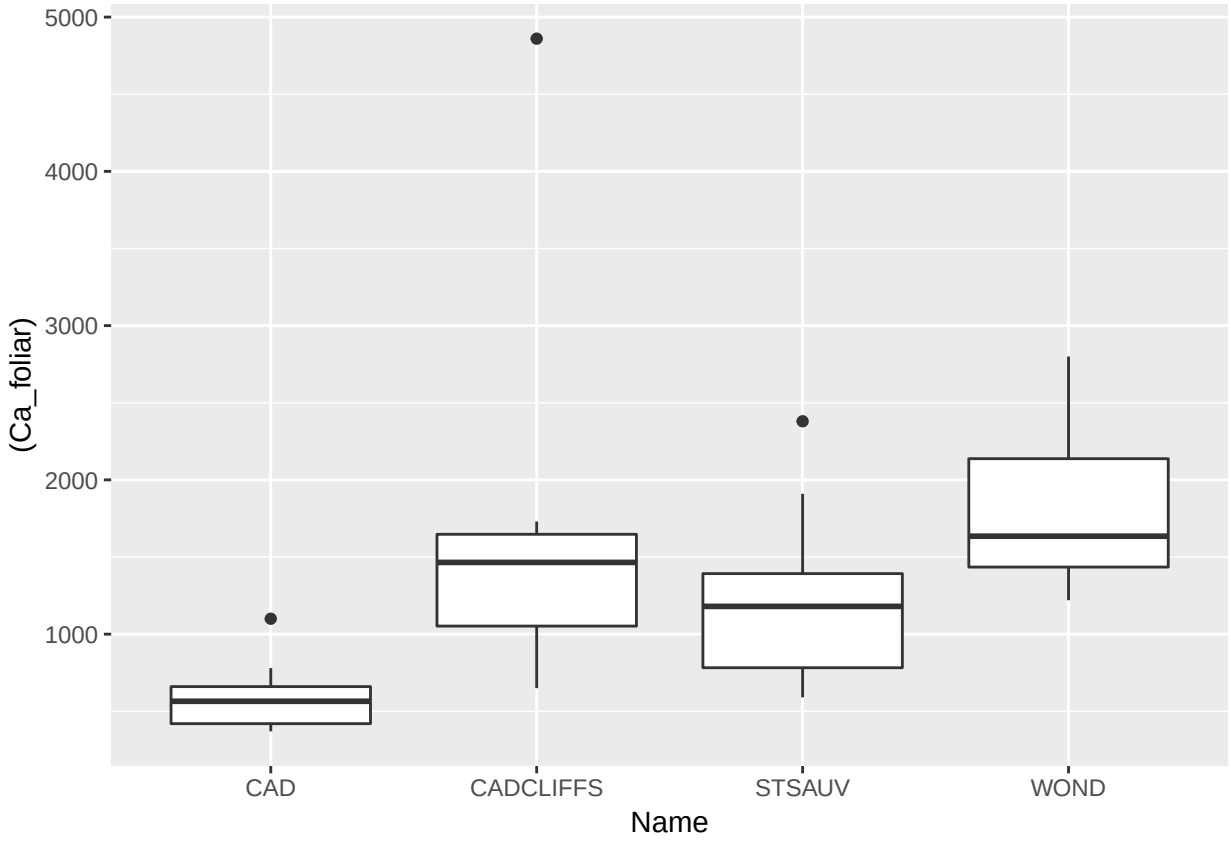
```
## Analysis of Variance Table
##
## Response: Ca_foliar
##                     Df   Sum Sq Mean Sq F value    Pr(>F)
## elevation_fac        1  6814502 6814502 13.0259 0.0009267 ***
## fire                 1  1556303 1556303  2.9749 0.0931449 .
## elevation_fac:fire   1   526703  526703  1.0068 0.3223708
## Residuals           36 18833470  523152
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
cld(emmeans(Ca_foliar_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean  SE df lower.CL upper.CL .group
##  high          fire       597 229 36      133     1061  1
##  high          no fire   1221 229 36      757     1685  12
##  low           fire      1652 229 36     1188     2116   2
##  low           no fire   1817 229 36     1353     2281   2
##
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (Ca_foliar))) +
  geom_boxplot()
```




```
### P_foliar
P_foliar_lm = lm(as.formula(paste(dep_variables[11],
                                  paste(ind_variables, collapse = "*"),
                                  sep = "~")), data = data)
#plot(resid(P_foliar_lm) ~ fitted(P_foliar_lm))
anova(P_foliar_lm)
```

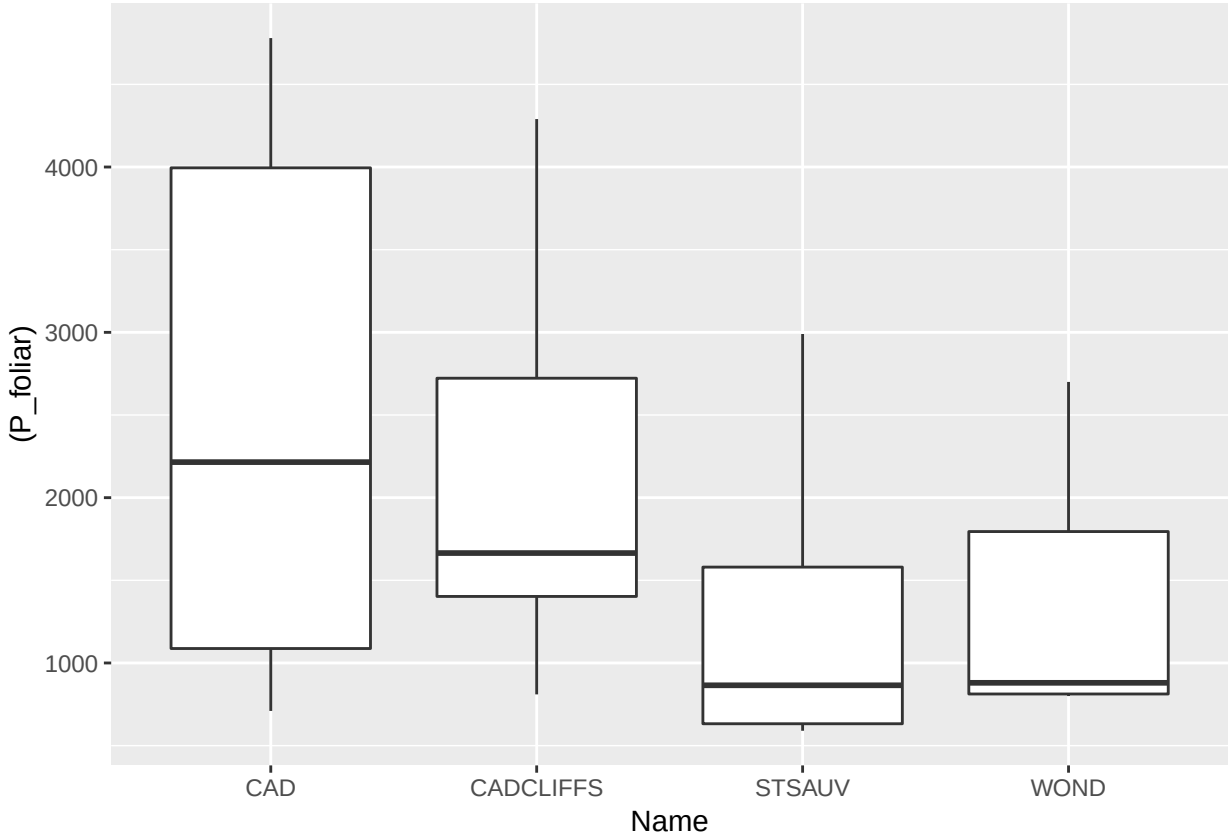
```
## Analysis of Variance Table
##
## Response: log(P_foliar)
##                    Df  Sum Sq Mean Sq F value   Pr(>F)
## elevation_fac       1  0.0296  0.0296  0.0796 0.779502
```

```
## fire                1  3.2582  3.2582  8.7713 0.005391 **
## elevation_fac:fire  1  0.0514  0.0514  0.1382 0.712210
## Residuals          36 13.3724  0.3715
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
cld(emmeans(P_foliar_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean    SE df lower.CL upper.CL .group
##  high          no fire   6.93 0.193 36     6.54     7.32  1
##  low           no fire   7.06 0.193 36     6.67     7.45  1
##  low           fire      7.56 0.193 36     7.17     7.95  1
##  high          fire      7.58 0.193 36     7.19     7.97  1
##
## Results are given on the log (not the response) scale.
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (P_foliar))) +
  geom_boxplot()
```



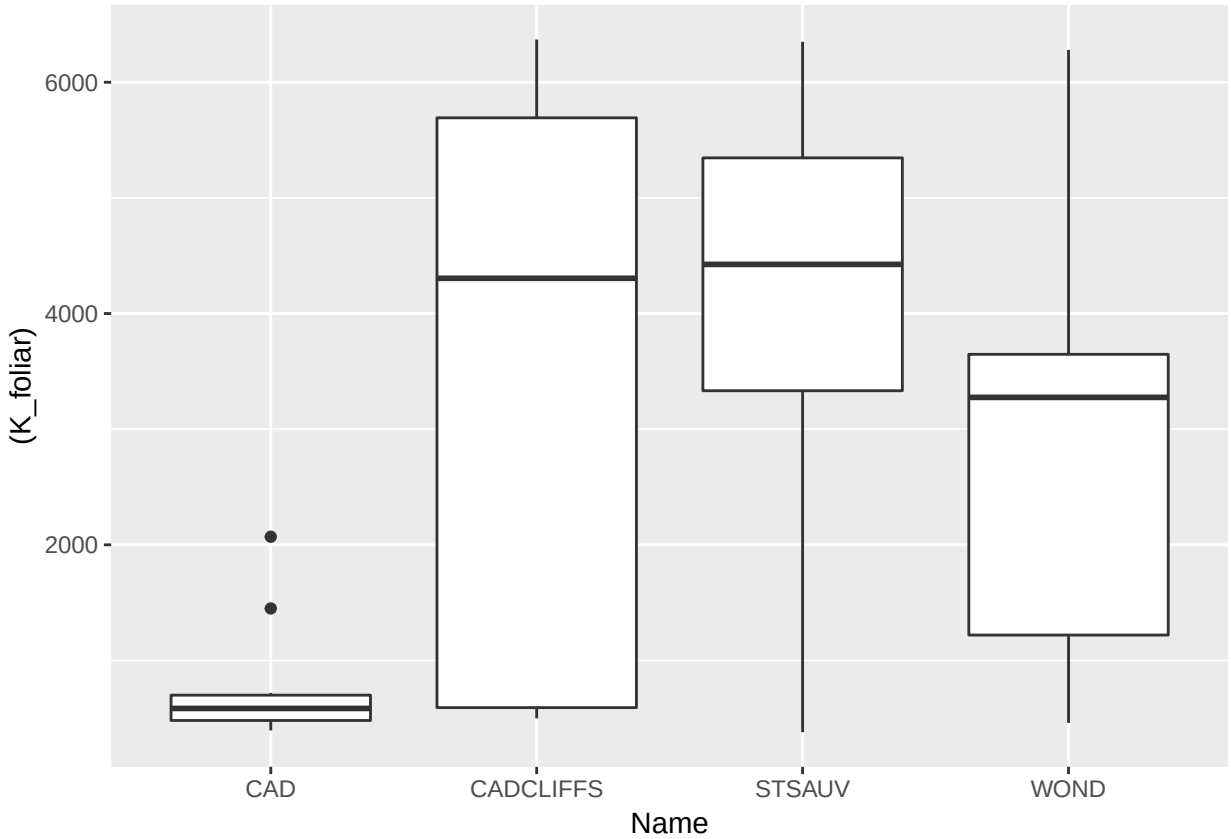

```
### K_foliar
K_foliar_lm = lm(as.formula(paste(dep_variables[12],
                                  paste(ind_variables, collapse = "*"),
                                  sep = "~")), data = data)
#plot(resid(K_foliar_lm) ~ fitted(K_foliar_lm))
anova(K_foliar_lm)
```

```
## Analysis of Variance Table
##
## Response: log(K_foliar)
##                    Df  Sum Sq Mean Sq F value  Pr(>F)
## elevation_fac       1  1.6363  1.6363  1.8751 0.17938
## fire                1  5.2363  5.2363  6.0005 0.01930 *
## elevation_fac:fire  1  5.6572  5.6572  6.4828 0.01531 *
## Residuals          36 31.4152  0.8726
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
cld(emmeans(K_foliar_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean    SE df lower.CL upper.CL .group
##  high          fire      6.52 0.295 36     5.92     7.12  1
##  low           no fire   7.65 0.295 36     7.05     8.25   2
##  low           fire      7.68 0.295 36     7.08     8.28   2
##  high          no fire   8.00 0.295 36     7.40     8.60   2
##
## Results are given on the log (not the response) scale.
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (K_foliar))) +
  geom_boxplot()
```




```
### Mg_foliar
Mg_foliar_lm = lm(as.formula(paste(dep_variables[13],
```

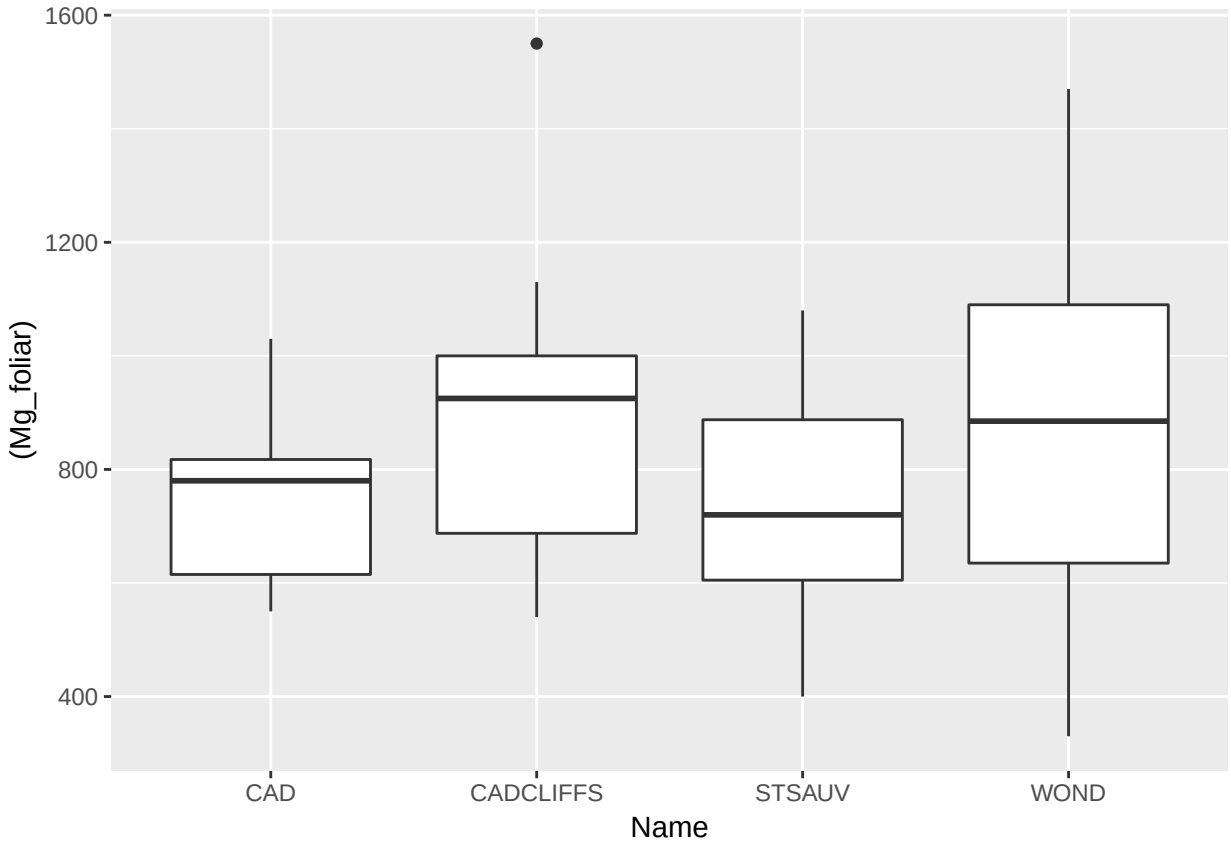
```
                                    paste(ind_variables, collapse = "*"),
                                    sep = "~")), data = data)
#plot(resid(Mg_foliar_lm) ~ fitted(Mg_foliar_lm))
anova(Mg_foliar_lm)
```

```
## Analysis of Variance Table
##
## Response: Mg_foliar
##                     Df  Sum Sq Mean Sq F value  Pr(>F)
## elevation_fac        1  231040  231040  3.0242 0.09058 .
## fire                 1   11560   11560  0.1513 0.69957
## elevation_fac:fire   1      90      90  0.0012 0.97281
## Residuals           36 2750300   76397
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
cld(emmeans(Mg_foliar_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean   SE df lower.CL upper.CL .group
##  high          no fire    725 87.4 36      548      902  1
##  high          fire       762 87.4 36      585      939  1
##  low           no fire    880 87.4 36      703     1057  1
##  low           fire       911 87.4 36      734     1088  1
##
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (Mg_foliar))) +
  geom_boxplot()
```



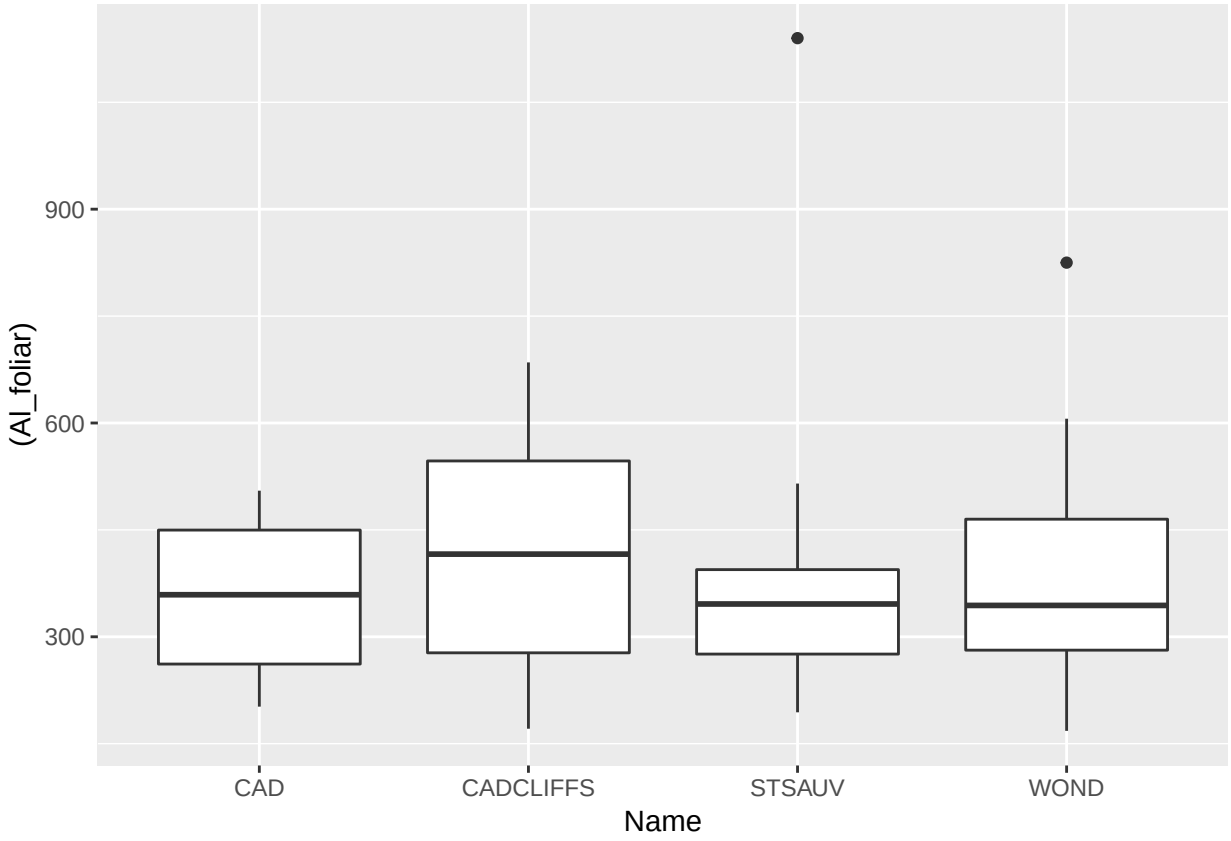

```
### Al_foliar
Al_foliar_lm = lm(as.formula(paste(dep_variables[14],
                                   paste(ind_variables, collapse = "*"),
                                   sep = "~")), data = data)
#plot(resid(Al_foliar_lm) ~ fitted(Al_foliar_lm))
anova(Al_foliar_lm)
```

```
## Analysis of Variance Table
##
## Response: Al_foliar
##                     Df  Sum Sq Mean Sq F value Pr(>F)
## elevation_fac        1    4928    4928  0.1266 0.7241
## fire                 1    2856    2856  0.0733 0.7881
## elevation_fac:fire   1   14440   14440  0.3708 0.5464
## Residuals           36 1401817   38939
```

```
cld(emmeans(Al_foliar_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean   SE df lower.CL upper.CL .group
##  high          fire       353 62.4 36      227      480  1
##  low           no fire    392 62.4 36      266      519  1
##  high          no fire    408 62.4 36      282      535  1
##  low           fire       414 62.4 36      287      540  1
##
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (Al_foliar))) +
  geom_boxplot()
```




```
### Zn_foliar
Zn_foliar_lm = lm(as.formula(paste(dep_variables[15],
                                   paste(ind_variables, collapse = "*"),
                                   sep = "~")), data = data)
#plot(resid(Zn_foliar_lm) ~ fitted(Zn_foliar_lm))
anova(Zn_foliar_lm)
```

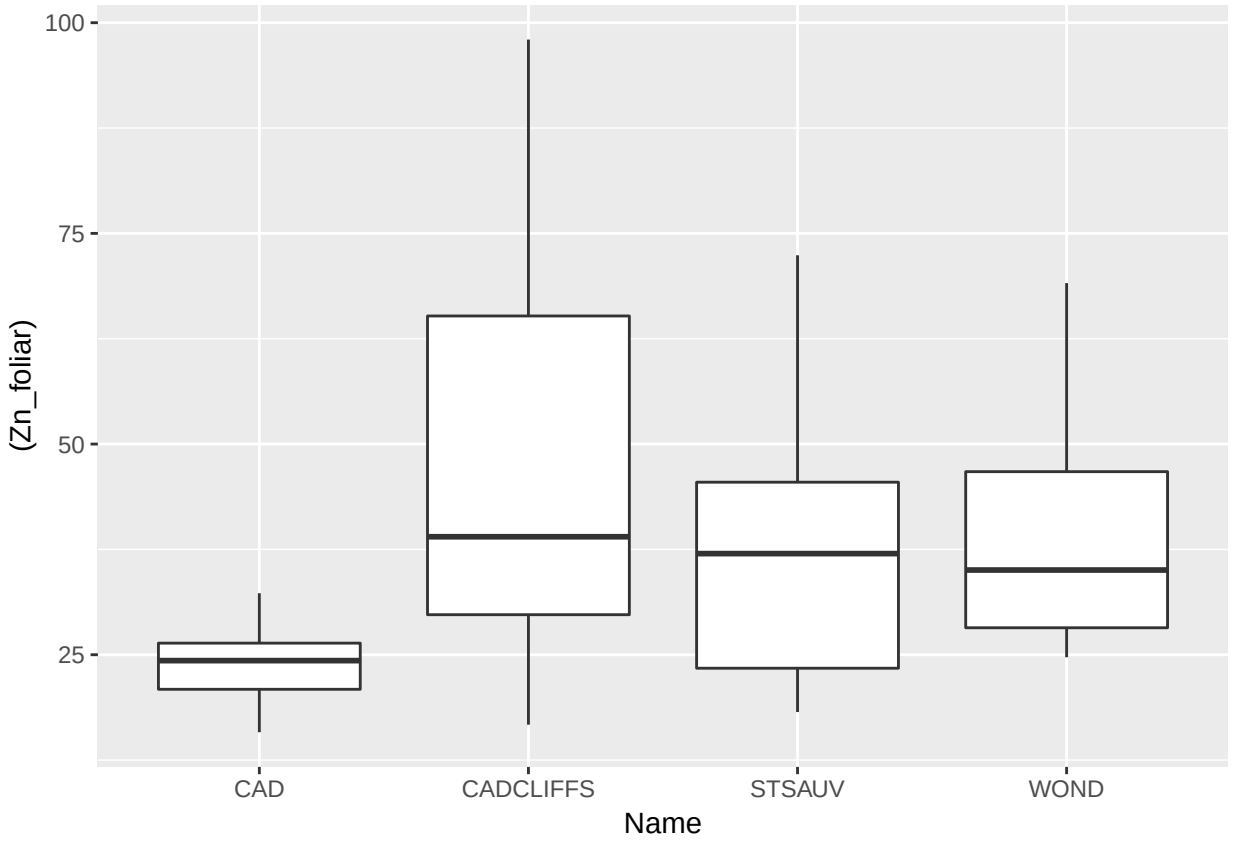
```
## Analysis of Variance Table
##
## Response: log(Zn_foliar)
##                     Df Sum Sq Mean Sq F value  Pr(>F)
## elevation_fac        1 1.0906 1.09060  6.7570 0.01345 *
## fire                 1 0.1143 0.11428  0.7080 0.40565
## elevation_fac:fire   1 0.6502 0.65022  4.0286 0.05229 .
## Residuals           36 5.8105 0.16140
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
cld(emmeans(Zn_foliar_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean    SE df lower.CL upper.CL .group
##  high          fire      3.16 0.127 36     2.90     3.42  1
##  high          no fire   3.52 0.127 36     3.27     3.78  12
##  low           no fire   3.60 0.127 36     3.34     3.86  12
##  low           fire      3.75 0.127 36     3.49     4.01   2
```

```
##
## Results are given on the log (not the response) scale.
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (Zn_foliar))) +
  geom_boxplot()
```




```
### Ca_soil
Ca_soil_lm = lm(as.formula(paste(dep_variables[16],
                                 paste(ind_variables, collapse = "*"),
                                 sep = "~")), data = data)
#plot(resid(Ca_soil_lm) ~ fitted(Ca_soil_lm))
anova(Ca_soil_lm)
```

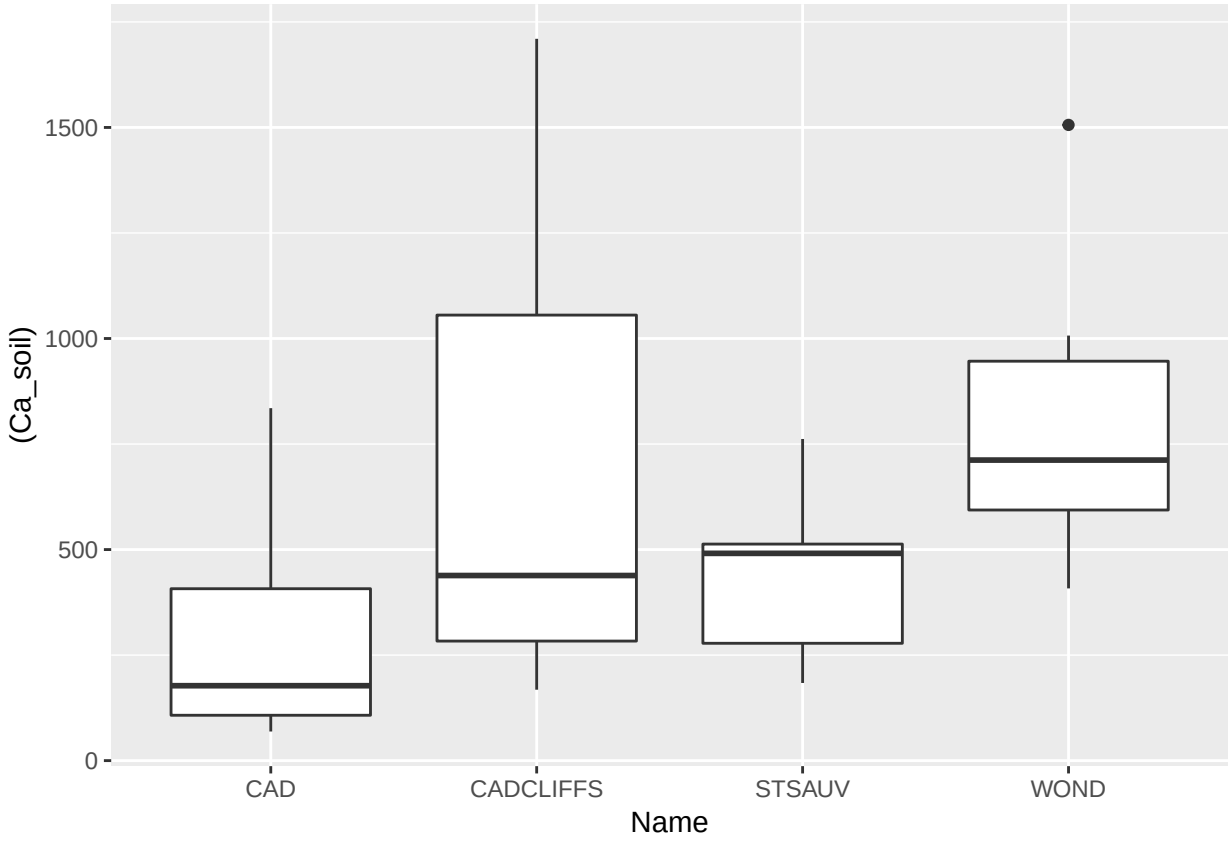
```
## Analysis of Variance Table
##
## Response: Ca_soil
##                    Df  Sum Sq Mean Sq F value  Pr(>F)
## elevation_fac       1 1131016 1131016  7.4816 0.01088 *
## fire                1   88710   88710  0.5868 0.45030
## elevation_fac:fire  1    1116    1116  0.0074 0.93218
## Residuals          27 4081653  151172
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
cld(emmeans(Ca_soil_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean  SE df lower.CL upper.CL .group
##  high          fire       312 137 27     29.7      594  1
##  high          no fire    431 147 27    129.8      733  1
##  low           fire       702 137 27    419.9      984  1
##  low           no fire    798 137 27    515.4     1080  1
##
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (Ca_soil))) +
  geom_boxplot()
```

```
## Warning: Removed 9 rows containing non-finite values (stat_boxplot).
```




```
### P_soil
P_soil_lm = lm(as.formula(paste(dep_variables[17],
                                paste(ind_variables, collapse = "*"),
                                sep = "~")), data = data)
#plot(resid(P_soil_lm) ~ fitted(P_soil_lm))
anova(P_soil_lm)
```

```
## Analysis of Variance Table
##
## Response: log(P_soil)
##                    Df  Sum Sq Mean Sq F value  Pr(>F)
## elevation_fac       1  3.5797  3.5797  3.2947 0.08063 .
## fire                1  2.1276  2.1276  1.9582 0.17309
## elevation_fac:fire  1  0.0072  0.0072  0.0066 0.93593
```

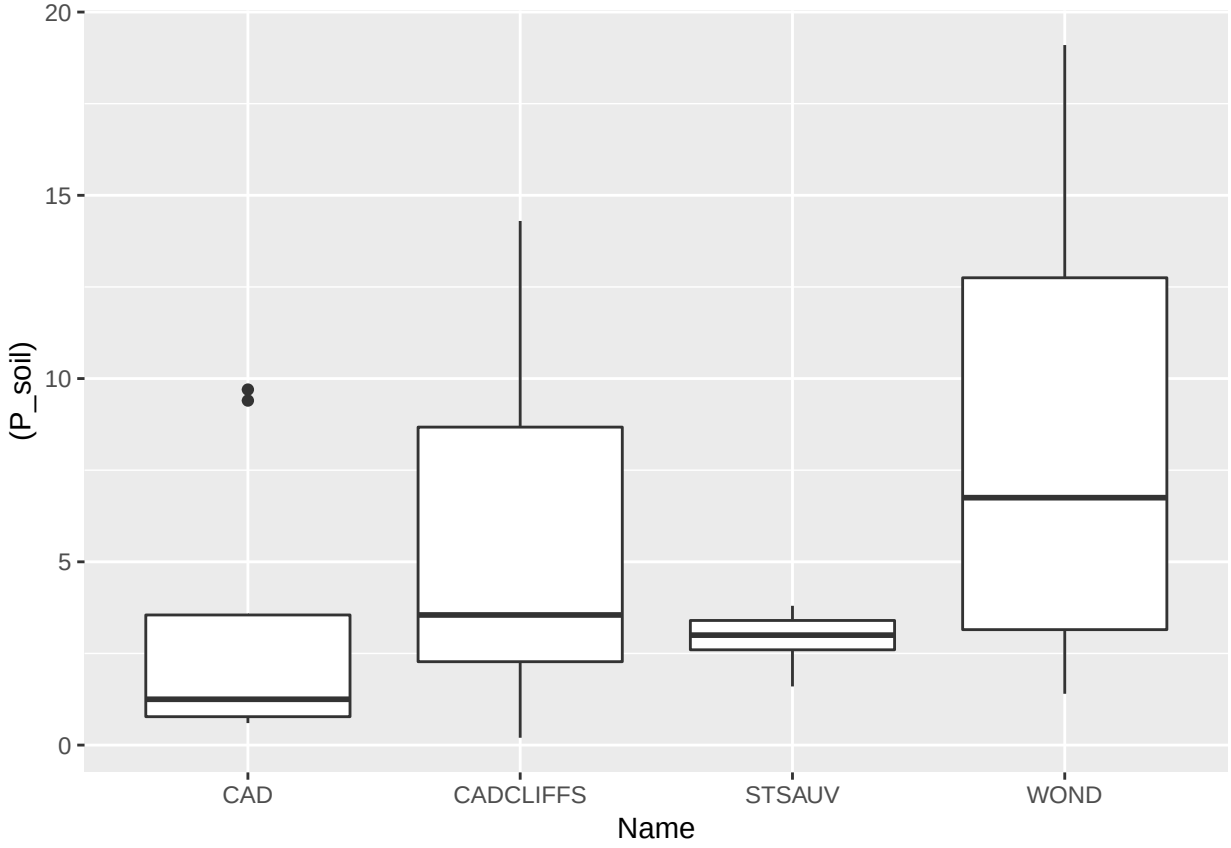
```
## Residuals          27 29.3360  1.0865
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
cld(emmeans(P_soil_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean    SE df lower.CL upper.CL .group
##  high          fire      0.54 0.369 27   -0.216     1.30  1
##  high          no fire   1.03 0.394 27    0.225     1.84  1
##  low           fire      1.17 0.369 27    0.417     1.93  1
##  low           no fire   1.73 0.369 27    0.971     2.48  1
##
## Results are given on the log (not the response) scale.
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (P_soil))) +
  geom_boxplot()
```

```
## Warning: Removed 9 rows containing non-finite values (stat_boxplot).
```




```
### K_soil
K_soil_lm = lm(as.formula(paste(dep_variables[18],
                                paste(ind_variables, collapse = "*"),
                                sep = "~")), data = data)
#plot(resid(K_soil_lm) ~ fitted(K_soil_lm))
anova(K_soil_lm)
```

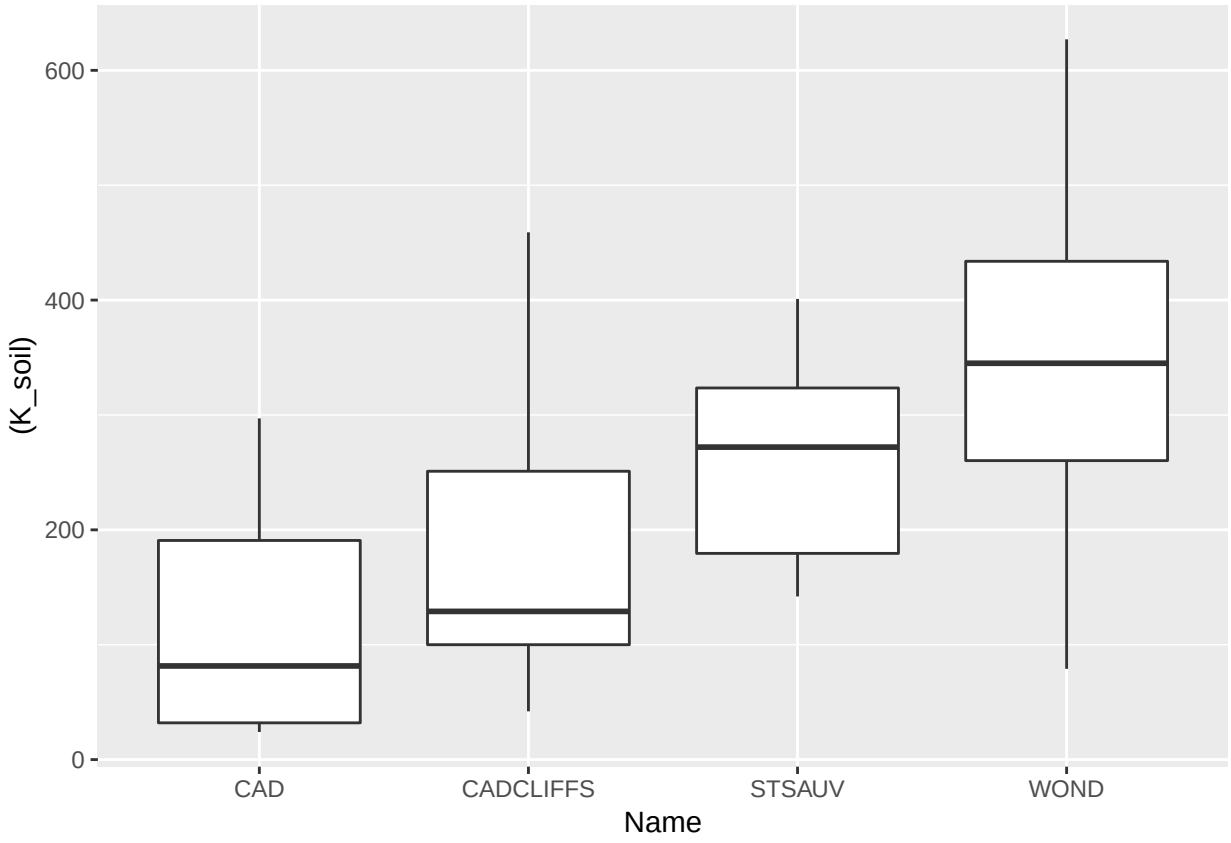
```
## Analysis of Variance Table
##
## Response: K_soil
##                    Df Sum Sq Mean Sq F value   Pr(>F)
## elevation_fac       1  51608   51608  2.7943 0.106149
## fire                1 164484  164484  8.9061 0.005971 **
## elevation_fac:fire  1    470     470  0.0255 0.874394
## Residuals          27 498657   18469
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
cld(emmeans(K_soil_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean   SE df lower.CL upper.CL .group
##  high          fire       122 48.0 27     23.8      221  1
##  low           fire       192 48.0 27     93.0      290  12
##  high          no fire    260 51.4 27    154.7      366  12
##  low           no fire    345 48.0 27    246.4      444   2
##
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (K_soil))) +
  geom_boxplot()
```

```
## Warning: Removed 9 rows containing non-finite values (stat_boxplot).
```



```r
### Mg_soil
Mg_soil_lm = lm(as.formula(paste(dep_variables[19],
                                 paste(ind_variables, collapse = "*"),
                                 sep = "~")), data = data)
#plot(resid(Mg_soil_lm) ~ fitted(Mg_soil_lm))
anova(Mg_soil_lm)
```

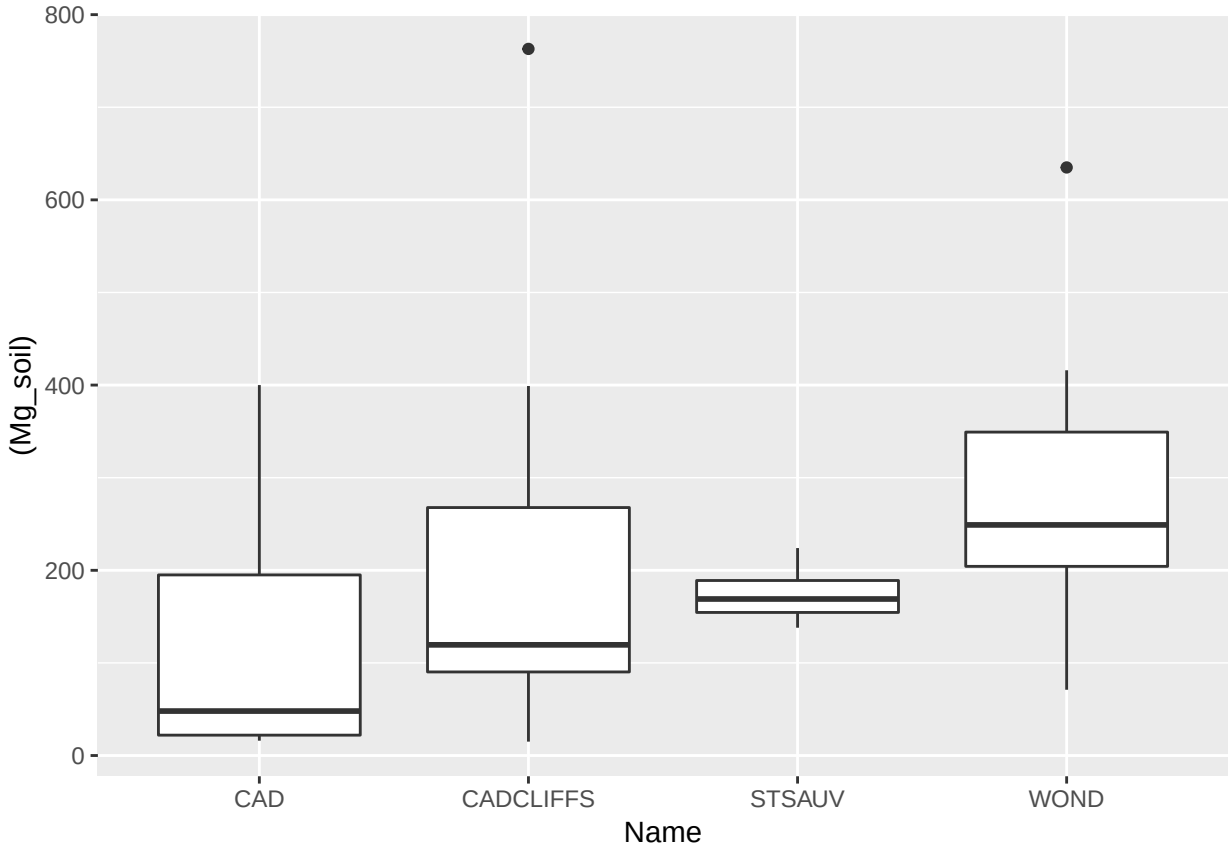
```
## Analysis of Variance Table
##
## Response: Mg_soil
##                     Df Sum Sq Mean Sq F value  Pr(>F)
## elevation_fac        1  89281   89281  2.9207 0.09892 .
## fire                 1  22475   22475  0.7352 0.39874
## elevation_fac:fire   1   1447    1447  0.0473 0.82939
## Residuals           27 825345   30568
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```r
cld(emmeans(Mg_soil_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean   SE df lower.CL upper.CL .group
##  high          fire       134 61.8 27     7.42      261  1
##  high          no fire    174 66.1 27    38.41      310  1
##  low           fire       227 61.8 27    99.79      353  1
##  low           no fire    294 61.8 27   166.92      421  1
##
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```r
ggplot(data = data, aes(x = Name, y = (Mg_soil))) +
  geom_boxplot()
```

```
## Warning: Removed 9 rows containing non-finite values (stat_boxplot).
```




```
### Al_soil
Al_soil_lm = lm(as.formula(paste(dep_variables[20],
                                 paste(ind_variables, collapse = "*"),
                                 sep = "~")), data = data)
#plot(resid(Al_soil_lm) ~ fitted(Al_soil_lm))
anova(Al_soil_lm)
```

```
## Analysis of Variance Table
##
## Response: log(Al_soil)
##                    Df Sum Sq Mean Sq F value   Pr(>F)
## elevation_fac       1 0.1771  0.1771  0.5440 0.467135
## fire                1 0.0021  0.0021  0.0065 0.936542
## elevation_fac:fire  1 3.3022  3.3022 10.1414 0.003637 **
## Residuals          27 8.7915  0.3256
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

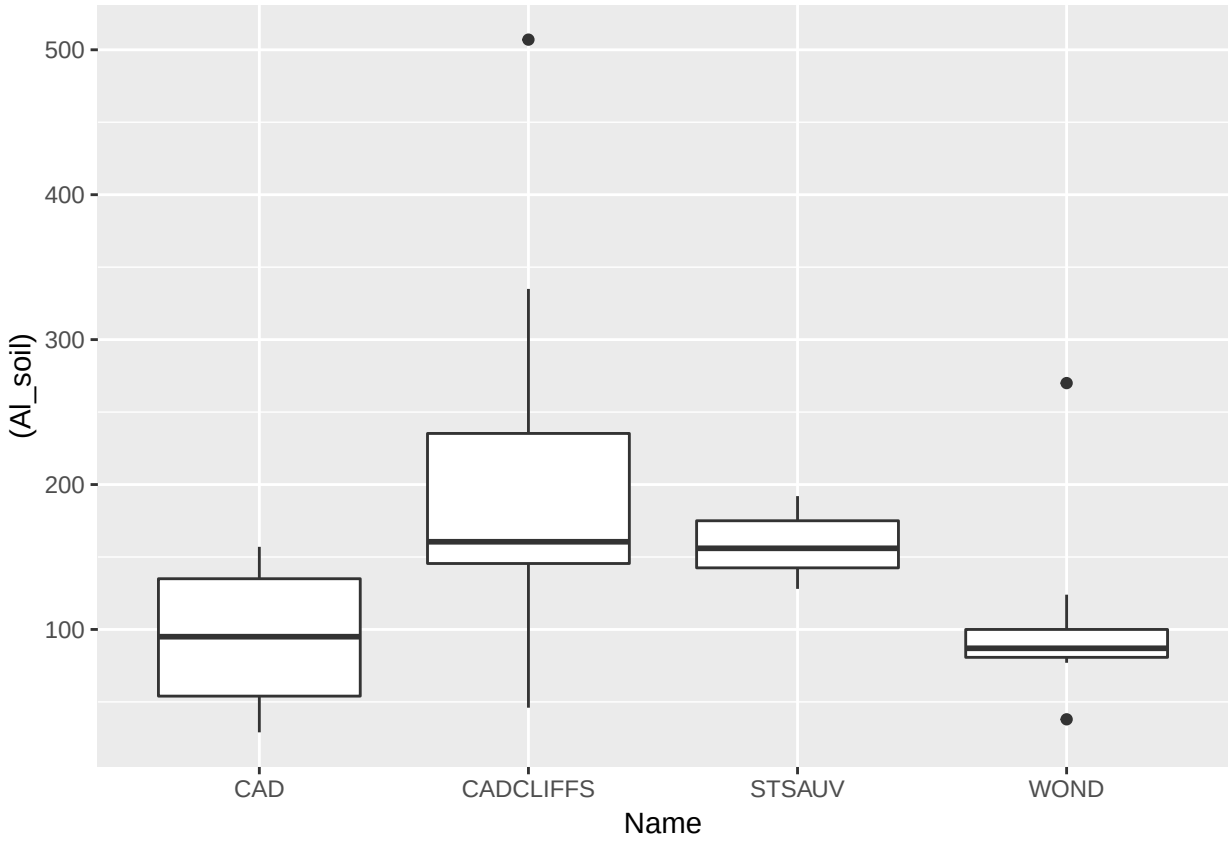
```
cld(emmeans(Al_soil_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean    SE df lower.CL upper.CL .group
##  high          fire      4.37 0.202 27     3.95     4.78  1
##  low           no fire   4.53 0.202 27     4.12     4.95  12
##  high          no fire   5.06 0.216 27     4.62     5.50  12
##  low           fire      5.15 0.202 27     4.73     5.56   2
##
## Results are given on the log (not the response) scale.
## Confidence level used: 0.95
```

```
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (Al_soil))) +
  geom_boxplot()
```

```
## Warning: Removed 9 rows containing non-finite values (stat_boxplot).
```




```
### Zn_soil
Zn_soil_lm = lm(as.formula(paste(dep_variables[21],
                                 paste(ind_variables, collapse = "*"),
                                 sep = "~")), data = data)
#plot(resid(Zn_soil_lm) ~ fitted(Zn_soil_lm))
anova(Zn_soil_lm)
```

```
## Analysis of Variance Table
##
## Response: log(Zn_soil)
##                    Df  Sum Sq Mean Sq F value  Pr(>F)
## elevation_fac       1  0.7821 0.78207  1.0531 0.31389
## fire                1  0.0005 0.00051  0.0007 0.97934
## elevation_fac:fire  1  2.6387 2.63871  3.5533 0.07023 .
## Residuals          27 20.0504 0.74261
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

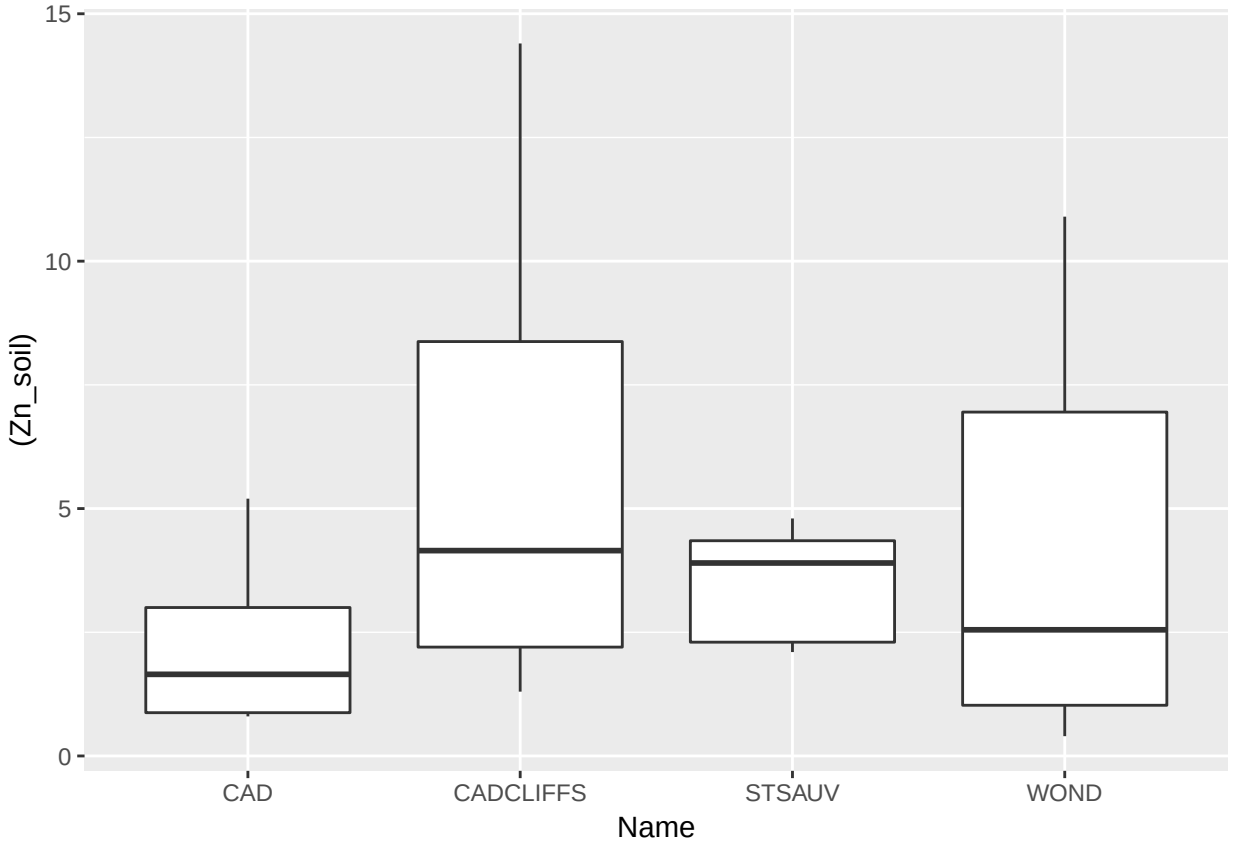
```
cld(emmeans(Zn_soil_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean    SE df lower.CL upper.CL .group
```

```
##  high          fire     0.569 0.305 27  -0.0561     1.19  1
##  low           no fire  0.895 0.305 27   0.2696     1.52  1
##  high          no fire  1.182 0.326 27   0.5135     1.85  1
##  low           fire     1.451 0.305 27   0.8258     2.08  1
##
## Results are given on the log (not the response) scale.
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (Zn_soil))) +
  geom_boxplot()
```

```
## Warning: Removed 9 rows containing non-finite values (stat_boxplot).
```




```
### pH
pH_lm = lm(as.formula(paste(dep_variables[22],
                            paste(ind_variables, collapse = "*"),
                            sep = "~")), data = data)
#plot(resid(pH_lm) ~ fitted(pH_lm))
anova(pH_lm)
```

```
## Analysis of Variance Table
##
## Response: pH
##                    Df  Sum Sq Mean Sq F value  Pr(>F)
## elevation_fac       1 0.36905 0.36905  3.6059 0.06831 .
## fire                1 0.00278 0.00278  0.0272 0.87028
```

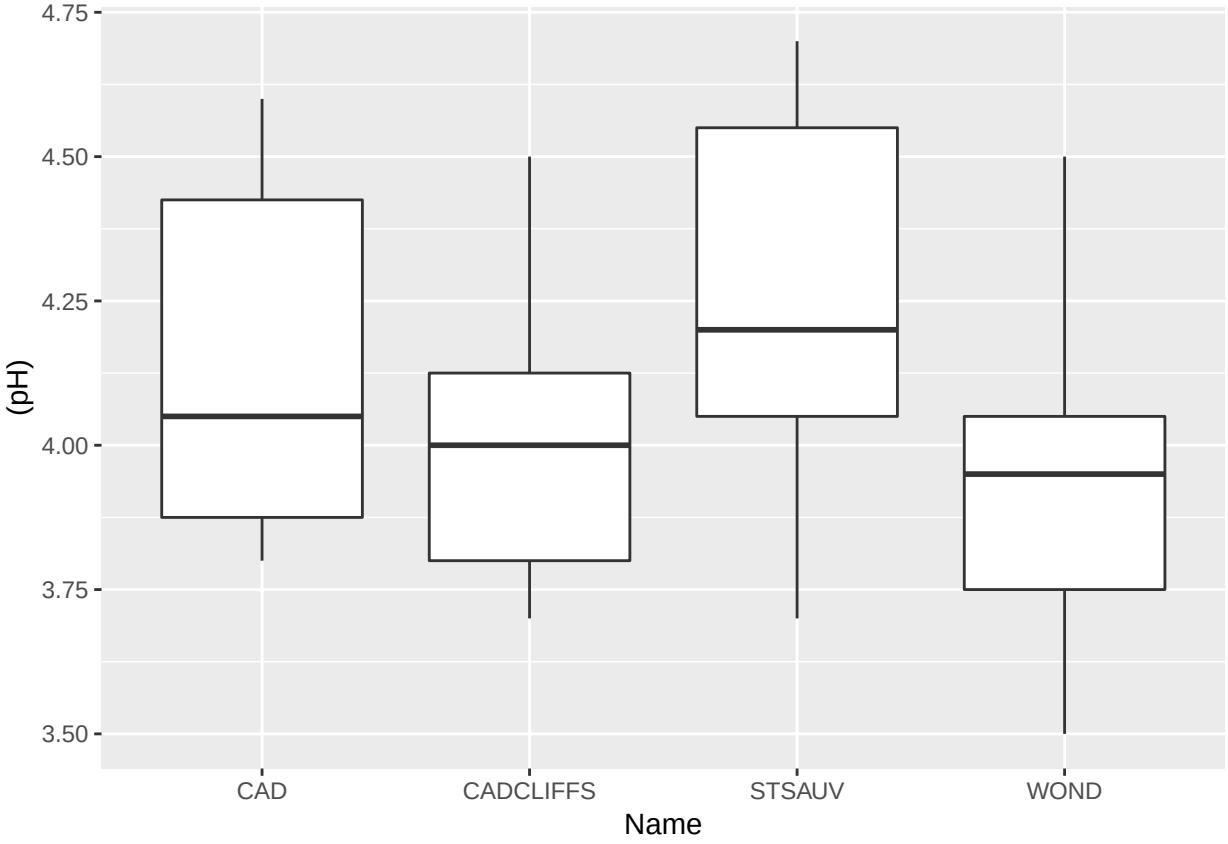
```
## elevation_fac:fire  1 0.07316 0.07316  0.7148 0.40528
## Residuals          27 2.76339 0.10235
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
cld(emmeans(pH_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean    SE df lower.CL upper.CL .group
##  low           no fire   3.94 0.113 27     3.71     4.17  1
##  low           fire      4.01 0.113 27     3.78     4.24  1
##  high          fire      4.14 0.113 27     3.91     4.37  1
##  high          no fire   4.26 0.121 27     4.01     4.51  1
##
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (pH))) +
  geom_boxplot()
```

```
## Warning: Removed 9 rows containing non-finite values (stat_boxplot).
```




```
### CEC
CEC_lm = lm(as.formula(paste(dep_variables[23],
                             paste(ind_variables, collapse = "*"),
                             sep = "~")), data = data)
#plot(resid(CEC_lm) ~ fitted(CEC_lm))
anova(CEC_lm)
```

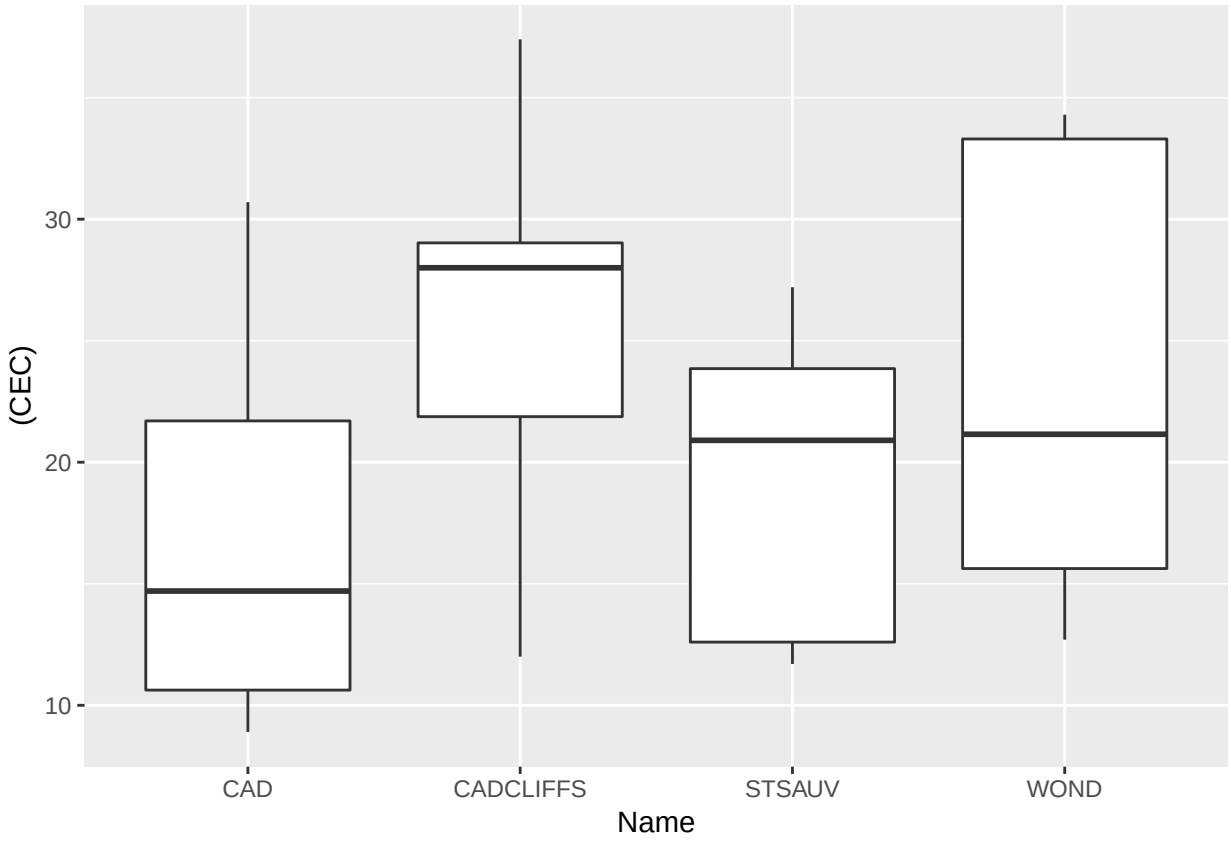
```
## Analysis of Variance Table
##
## Response: CEC
##                     Df  Sum Sq Mean Sq F value  Pr(>F)
## elevation_fac        1  318.02  318.02  4.8714 0.03599 *
## fire                 1    0.73    0.73  0.0112 0.91666
## elevation_fac:fire   1   28.92   28.92  0.4429 0.51136
## Residuals           27 1762.65   65.28
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
cld(emmeans(CEC_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean   SE df lower.CL upper.CL .group
##  high          fire      17.3 2.86 27     11.4     23.1  1
##  high          no fire   19.0 3.05 27     12.7     25.2  1
##  low           no fire   23.4 2.86 27     17.5     29.2  1
##  low           fire      25.6 2.86 27     19.7     31.4  1
##
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (CEC))) +
  geom_boxplot()
```

```
## Warning: Removed 9 rows containing non-finite values (stat_boxplot).
```



```r
### C_soil
C_soil_lm = lm(as.formula(paste(dep_variables[24],
                                paste(ind_variables, collapse = "*"),
                                sep = "~")), data = data)
#plot(resid(C_soil_lm) ~ fitted(C_soil_lm))
anova(C_soil_lm)
```

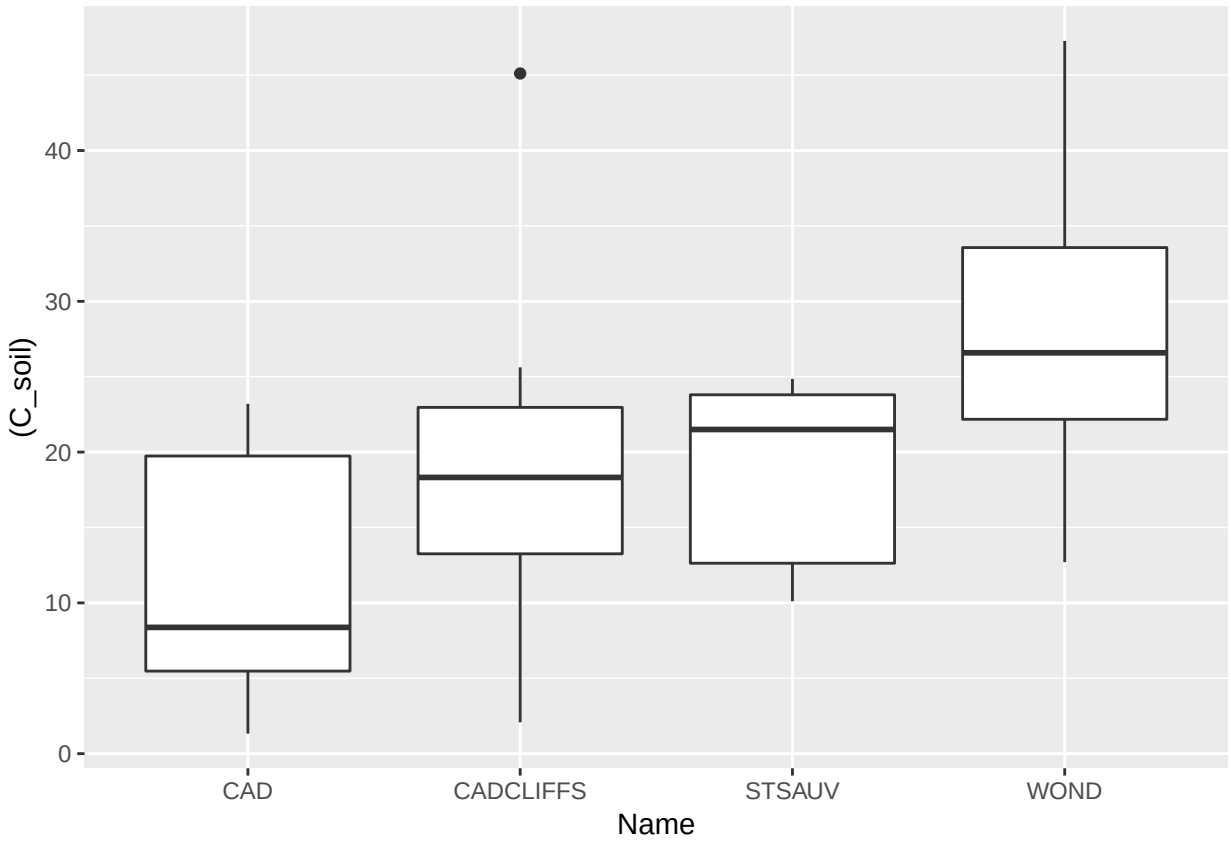
```
## Analysis of Variance Table
##
## Response: C_soil
##                     Df  Sum Sq Mean Sq F value  Pr(>F)
## elevation_fac        1  640.83  640.83  6.3631 0.01785 *
## fire                 1  480.00  480.00  4.7661 0.03789 *
## elevation_fac:fire   1    4.90    4.90  0.0487 0.82705
## Residuals           27 2719.18  100.71
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```r
cld(emmeans(C_soil_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean   SE df lower.CL upper.CL .group
##  high          fire      11.4 3.55 27     4.14     18.7  1
##  high          no fire   18.5 3.79 27    10.69     26.3  12
##  low           fire      19.5 3.55 27    12.21     26.8  12
##  low           no fire   28.1 3.55 27    20.85     35.4   2
##
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```r
ggplot(data = data, aes(x = Name, y = (C_soil))) +
  geom_boxplot()
```

```
## Warning: Removed 9 rows containing non-finite values (stat_boxplot).
```




```
### N_soil
N_soil_lm = lm(as.formula(paste(dep_variables[25],
                                paste(ind_variables, collapse = "*"),
                                sep = "~")), data = data)
#plot(resid(N_soil_lm) ~ fitted(N_soil_lm))
anova(N_soil_lm)
```

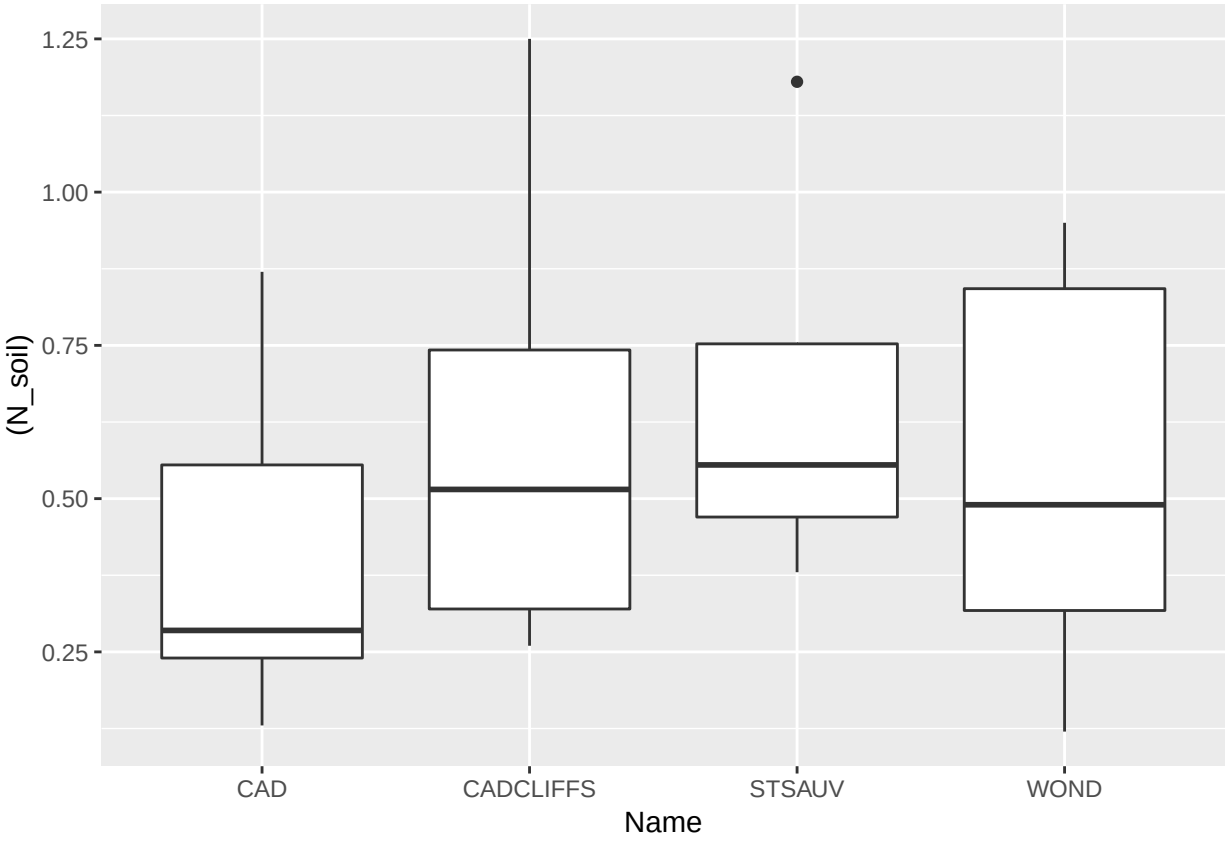
```
## Analysis of Variance Table
##
## Response: N_soil
##                    Df  Sum Sq  Mean Sq F value Pr(>F)
## elevation_fac       1 0.04513 0.045129  0.4138 0.5267
## fire                1 0.03916 0.039162  0.3591 0.5551
## elevation_fac:fire  1 0.15121 0.151209  1.3864 0.2516
## Residuals          22 2.39950 0.109068
```

```
cld(emmeans(N_soil_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean    SE df lower.CL upper.CL .group
##  high          fire     0.409 0.117 22    0.167    0.651  1
##  low           no fire  0.545 0.135 22    0.265    0.825  1
##  low           fire     0.604 0.117 22    0.362    0.846  1
##  high          no fire  0.667 0.165 22    0.325    1.010  1
##
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (N_soil))) +
  geom_boxplot()
```

```
## Warning: Removed 14 rows containing non-finite values (stat_boxplot).
```




```
### CN_soil
CN_soil_lm = lm(as.formula(paste(dep_variables[26],
                                 paste(ind_variables, collapse = "*"),
                                 sep = "~")), data = data)
#plot(resid(CN_soil_lm) ~ fitted(CN_soil_lm))
anova(CN_soil_lm)
```

```
## Analysis of Variance Table
##
## Response: log(CN_soil)
##                    Df Sum Sq Mean Sq F value  Pr(>F)
## elevation_fac       1 1.5999 1.59986  5.0654 0.03474 *
## fire                1 0.8234 0.82341  2.6070 0.12064
## elevation_fac:fire  1 0.6937 0.69367  2.1962 0.15253
## Residuals          22 6.9485 0.31584
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

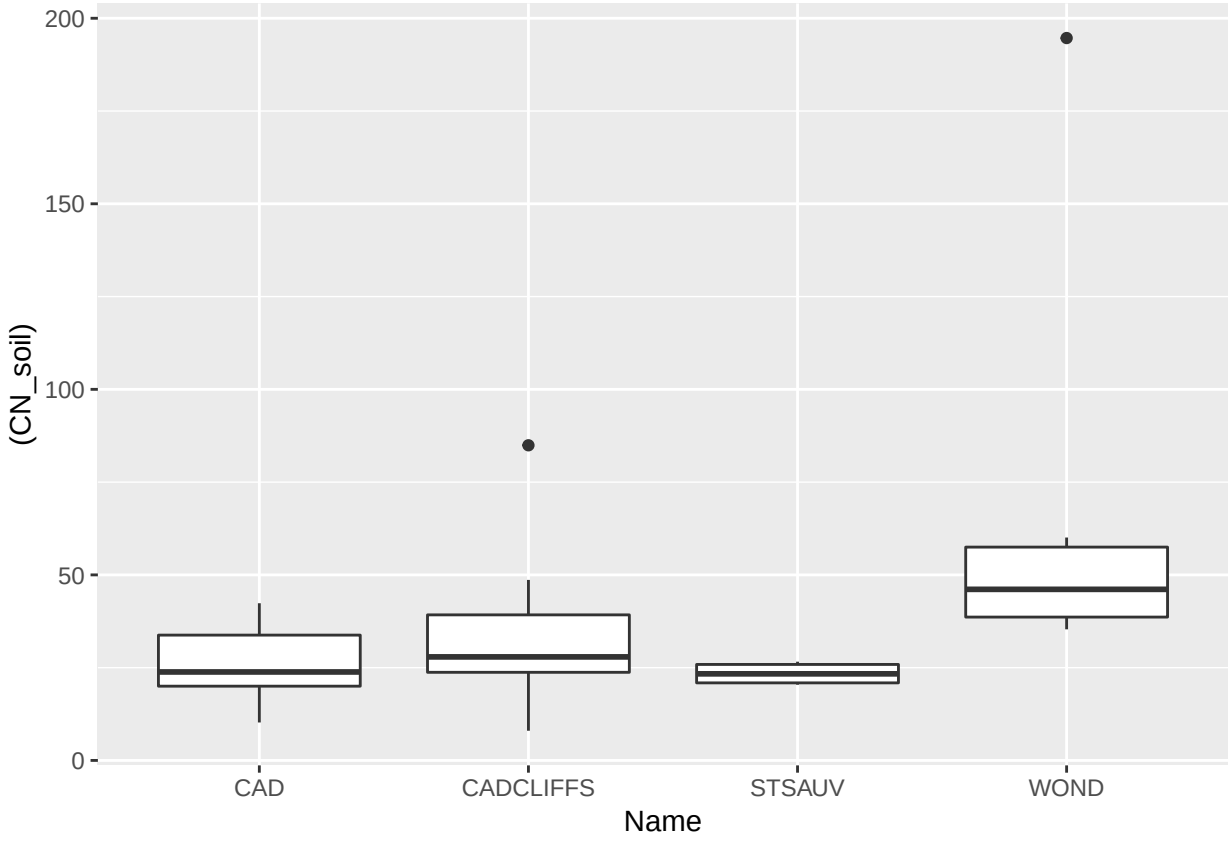
```
cld(emmeans(CN_soil_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean    SE df lower.CL upper.CL .group
##  high          no fire   3.15 0.281 22     2.56     3.73  12
##  high          fire      3.16 0.199 22     2.75     3.57  1
```

```
##  low           fire      3.37 0.199 22     2.96     3.78  12
##  low           no fire   4.03 0.229 22     3.56     4.51   2
##
## Results are given on the log (not the response) scale.
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (CN_soil))) +
  geom_boxplot()
```

```
## Warning: Removed 14 rows containing non-finite values (stat_boxplot).
```




```
### retention
retention_lm = lm(as.formula(paste(dep_variables[27],
                                   paste(ind_variables, collapse = "*"),
                                   sep = "~")), data = data)
#plot(resid(retention_lm) ~ fitted(retention_lm))
anova(retention_lm)
```

```
## Analysis of Variance Table
##
## Response: asin(sqrt(0.01 * retention))
##                    Df   Sum Sq  Mean Sq F value    Pr(>F)
## elevation_fac       1 0.006536 0.006536  0.8661 0.3582460
## fire                1 0.084133 0.084133 11.1488 0.0019655 **
## elevation_fac:fire  1 0.128378 0.128378 17.0119 0.0002093 ***
## Residuals          36 0.271670 0.007546
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
cld(emmeans(retention_lm, ~elevation_fac * fire))
```

```
##  elevation_fac fire    emmean     SE df lower.CL upper.CL .group
##  low           no fire  0.470 0.0275 36    0.414    0.526  1
##  high          fire     0.536 0.0275 36    0.480    0.592  1
##  high          no fire  0.558 0.0275 36    0.502    0.613  1
##  low           fire     0.675 0.0275 36    0.619    0.731   2
##
## Results are given on the asin(sqrt(mu)) (not the response) scale.
## Confidence level used: 0.95
## P value adjustment: tukey method for comparing a family of 4 estimates
## significance level used: alpha = 0.05
```

```
ggplot(data = data, aes(x = Name, y = (retention))) +
  geom_boxplot()
```