

Stem density

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Packages and data

```
library(car)
library(tidyverse)
library(lme4)

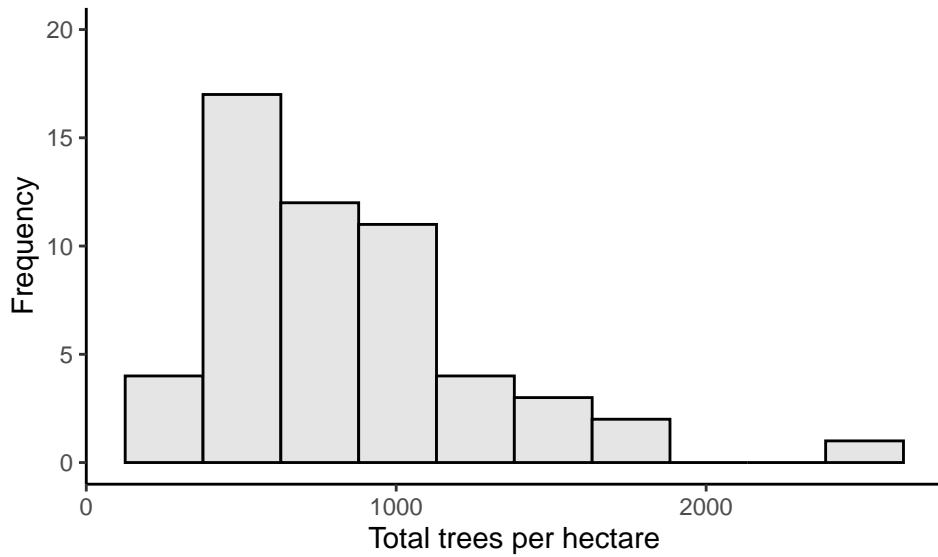
data_stems <- read_csv("data/data_stems.csv") %>%
  mutate(plot_id = as.factor(plot_id),
        size_class = as.factor(size_class),
        harvested = as.factor(harvested),
        vegetation_type = as.factor(vegetation_type),
        milpa = as.factor(milpa))
data_plots <- read_csv("data/data_plots.csv") %>%
  mutate(plot_id = as.factor(plot_id),
        harvested = as.factor(harvested),
        vegetation_type = as.factor(vegetation_type),
        milpa = as.factor(milpa))

plot_theme <- theme(plot.background = element_rect(fill = "white"),
                     plot.title = element_blank(),
                     plot.subtitle = element_text(family="sans", face="plain"),
                     axis.title.x = element_text(family="sans", face="bold"),
                     axis.title.y = element_text(family="sans", face="bold"),
                     axis.text.x = element_text(family="sans", face="plain"),
                     axis.text.y = element_text(family="sans", face="plain"),
                     panel.background = element_rect(fill="white"),
                     panel.grid.major.x = element_blank(),
                     panel.grid.major.y = element_line(color="gainsboro"),
                     panel.grid.minor = element_blank(),
                     axis.ticks = element_blank(),
                     legend.background = element_rect(color="black", fill = "white"),
                     legend.position = c(0.9, 0.85),
                     legend.title = element_text(color = "black", face = "bold", hjust = 0.5),
                     legend.text = element_text(color = "black"))
```

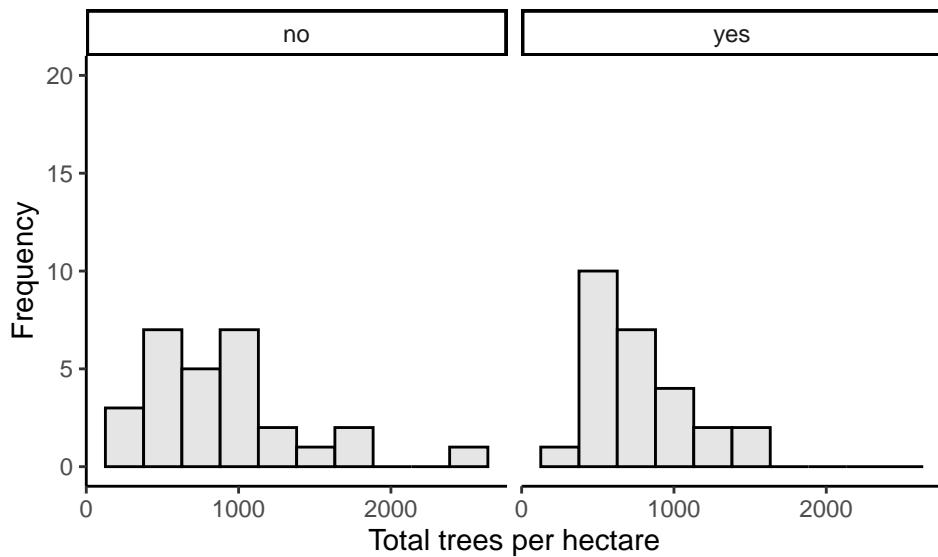
Figures

Histograms of total trees per hectare

```
# All plots
data_plots %>%
  ggplot() +
  geom_histogram(aes(x = stemden_totaltrees), bins = 10, fill = "gray90", col = "black") +
  theme_classic() + ylim(c(0,20)) + labs(x = "Total trees per hectare", y = "Frequency")
```



```
# Harvested vs. unharvested plots
data_plots %>%
  ggplot() +
  geom_histogram(aes(x = stemden_totaltrees), bins = 10, fill = "gray90", col = "black") +
  facet_wrap(~harvested) +
  theme_classic() + ylim(c(0,20)) + labs(x = "Total trees per hectare", y = "Frequency")
```



Five-number summaries of total trees per hectare

```
# All plots
summary(data_plots$stemden_totaltrees)

##      Min. 1st Qu. Median     Mean 3rd Qu.     Max.
##    254.6   557.0  779.9  848.8 1018.6 2514.6

# Harvested vs. unharvested plots
data_plots %>% filter(harvested == "yes") %>% pull(stemden_totaltrees) %>% summary()

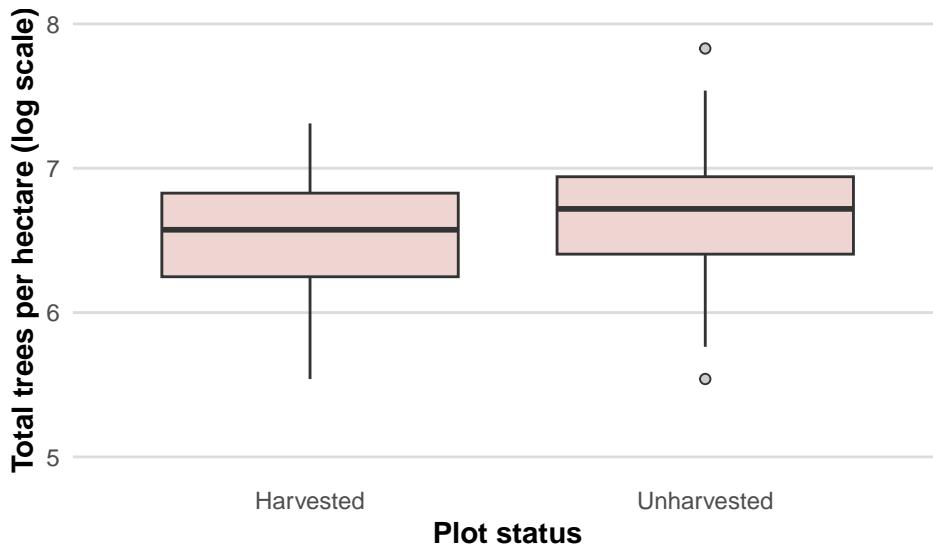
##      Min. 1st Qu. Median     Mean 3rd Qu.     Max.
##    254.6   517.3  716.2  778.6  923.1 1496.1

data_plots %>% filter(harvested == "no") %>% pull(stemden_totaltrees) %>% summary()

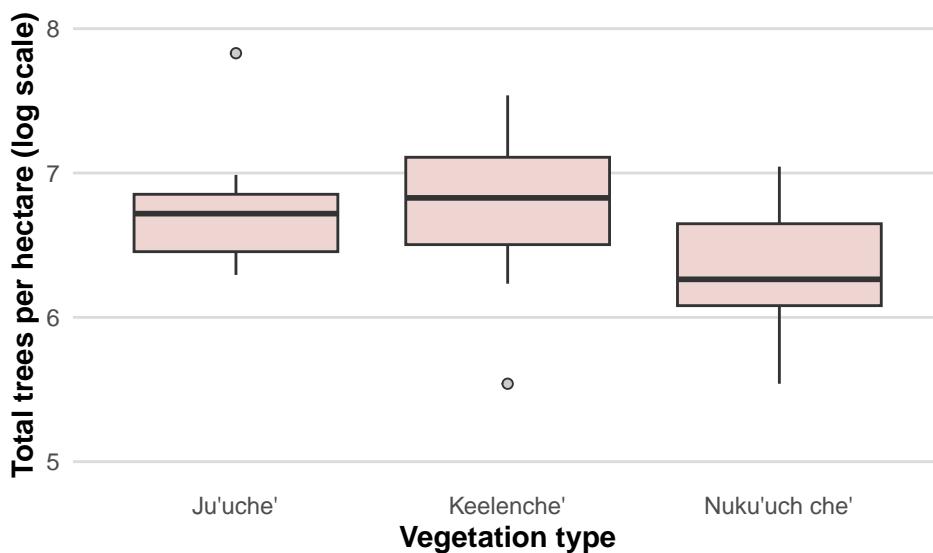
##      Min. 1st Qu. Median     Mean 3rd Qu.     Max.
##    254.6   604.8  827.6  914.0 1034.5 2514.6
```

Box plots of total trees per hectare (log scale)

```
# By harvesting status
stemden_fig1 <- data_plots %>%
  mutate(harvested = fct_recode(harvested,
                                "Harvested" = "yes", "Unharvested" = "no")) %>%
  ggplot() +
  geom_boxplot(aes(x = fct_relevel(harvested, "Harvested", "Unharvested"),
                   y = log(stemden_totaltrees)),
                fill = "mistyrose2", outlier.shape = 21,
                outlier.color = "gray20", outlier.fill = "gray80") +
  plot_theme +
  coord_cartesian(ylim = c(5, 8)) +
  labs(x = "Plot status", y = "Total trees per hectare (log scale)")
ggsave("figures/stemden_fig1.png", height = 4, width = 6)
stemden_fig1
```



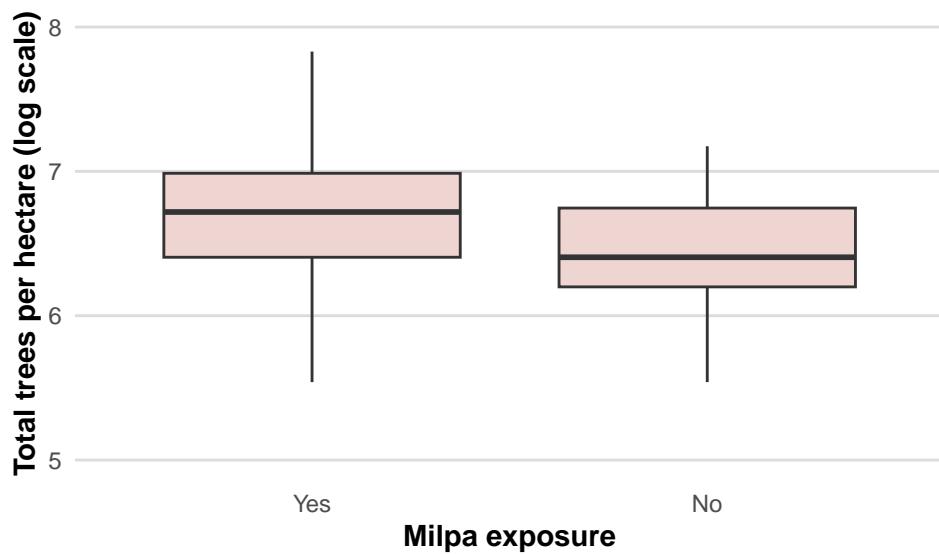
```
# By vegetation type
stemden_fig2 <- data_plots %>%
  mutate(vegetation_type = fct_recode(vegetation_type,
    "Ju'uche'" = "juuche",
    "Keelenche'" = "keelenche",
    "Nuku'uch che'" = "nukuuchche")) %>%
  ggplot() +
  geom_boxplot(aes(x = vegetation_type, y = log(stemden_totaltrees)),
    fill = "mistyrose2", outlier.shape = 21,
    outlier.color = "gray20", outlier.fill = "gray80") +
  plot_theme +
  coord_cartesian(ylim = c(5, 8)) +
  labs(x = "Vegetation type", y = "Total trees per hectare (log scale)")
ggsave("figures/stemden_fig2.png", height = 4, width = 6)
stemden_fig2
```



```

# By milpa exposure
stemden_fig3 <- data_plots %>%
  mutate(milpa = fct_recode(milpa,
                            "Yes" = "yes", "No" = "no")) %>%
  ggplot() +
  geom_boxplot(aes(x = fct_relevel(milpa, "Yes", "No"), y = log(stemden_totaltrees)),
                fill = "mistyrose2", outlier.shape = 21,
                outlier.color = "gray20", outlier.fill = "gray80") +
  plot_theme +
  coord_cartesian(ylim = c(5, 8)) +
  labs(x = "Milpa exposure", y = "Total trees per hectare (log scale)")
ggsave("figures/stemden_fig3.png", height = 4, width = 6)
stemden_fig3

```

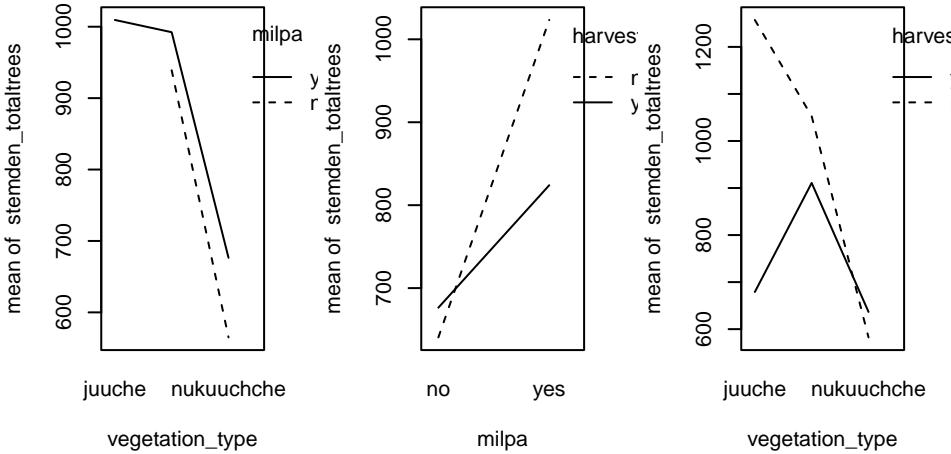


Interaction plots

```

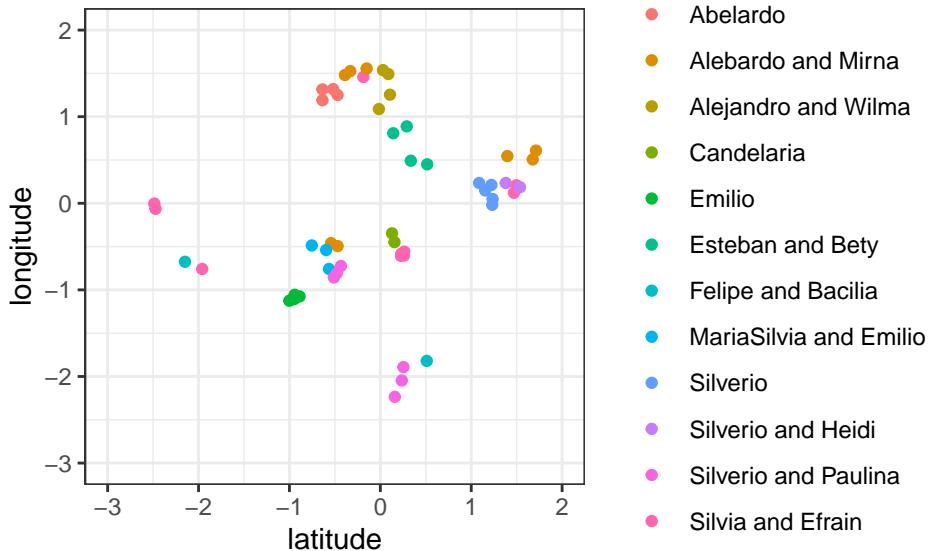
par(mfrow = c(1,3))
with(data_plots, {interaction.plot(vegetation_type, milpa, stemden_totaltrees)
                  interaction.plot(milpa, harvested, stemden_totaltrees)
                  interaction.plot(vegetation_type, harvested, stemden_totaltrees)})

```



Latitude and longitude by artisan

```
data_plots %>%
  ggplot(aes(x = latitude, y = longitude)) +
  geom_point(aes(color = artisan), shape = 19) +
  xlim(-3, 2) +
  ylim(-3, 2) +
  theme_bw()
```



Models

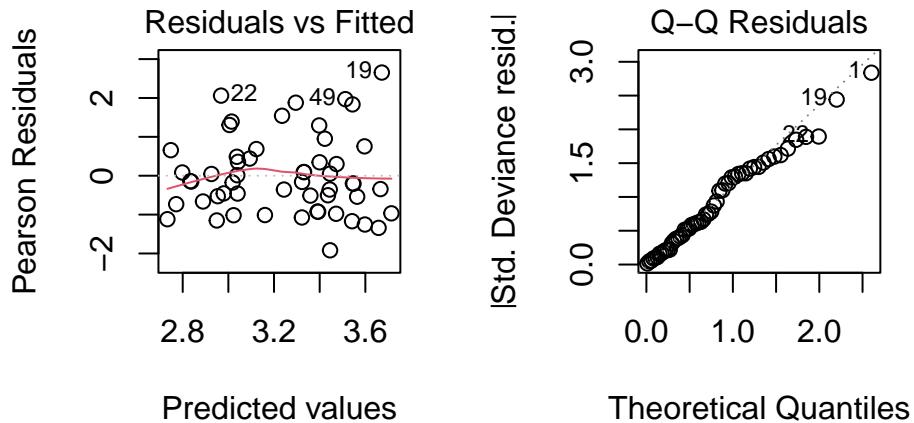
We have fewer than five observations for each artisan, so we elect not to include artisan as a random effect. Also, some individuals appear in more than one level of artisan, so we could not disentangle variability due to each individual.

We decide to fit a negative binomial model with all other covariates, including latitude and longitude. We found that the variance of the residuals increases quadratically with the mean, so we believe a negative binomial model is more appropriate than a quasipoisson GLM.

Fit negative binomial model and check residual plots

```
mod <- MASS::glm.nb(stemden_totaltrees_count ~ harvested + vegetation_type + milpa +
                     harvested:vegetation_type + harvested:milpa + latitude + longitude +
                     offset(plots_per_ha), data = data_plots)

par(mfrow = c(1,2))
plot(mod, which = 1:2)
```



Summarize the model

```
summary(mod)
```

```
##
## Call:
## MASS::glm.nb(formula = stemden_totaltrees_count ~ harvested +
##   vegetation_type + milpa + harvested:vegetation_type + harvested:milpa +
##   latitude + longitude + offset(plots_per_ha), data = data_plots,
##   init.theta = 8.413800108, link = log)
##
## Coefficients:
## (Intercept)             Estimate Std. Error z value Pr(>|z|)
## harvestedyes            3.406376  0.298190 11.424 <2e-16
## vegetation_typekeelenche -0.555248  0.414112 -1.341  0.1800
## vegetation_typepenukuuchche -0.179506  0.227138 -0.790  0.4294
## milpayes                 -0.595615  0.285526 -2.086  0.0370
## latitude                  0.238074  0.214022  1.112  0.2660
## longitude                 -0.004282  0.059109 -0.072  0.9423
## longitude                 -0.086717  0.071057 -1.220  0.2223
## harvestedyes:vegetation_typekeelenche  0.463946  0.339983  1.365  0.1724
## harvestedyes:vegetation_typepenukuuchche  0.680179  0.392460  1.733  0.0831
## harvestedyes:milpayes           -0.056634  0.282870 -0.200  0.8413
##
## (Intercept) ***
```

harvestedyes

```

## vegetation_typekeelenche
## vegetation_typenukuuchche *
## milpayes
## latitude
## longitude
## harvestedyes:vegetation_typekeelenche
## harvestedyes:vegetation_typenukuuchche .
## harvestedyes:milpayes
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for Negative Binomial(8.4138) family taken to be 1)
##
## Null deviance: 78.876 on 53 degrees of freedom
## Residual deviance: 53.599 on 44 degrees of freedom
## AIC: 420.64
##
## Number of Fisher Scoring iterations: 1
##
##
## Theta: 8.41
## Std. Err.: 2.10
##
## 2 x log-likelihood: -398.639

```

Construct analysis of deviance table

```

# Use type 2 SS since data are unbalanced)
Anova(mod, type = 2, test.statistic = "LR")

```

```

## Analysis of Deviance Table (Type II tests)
##
## Response: stemden_totaltrees_count
##          LR Chisq Df Pr(>Chisq)
## harvested           1.1297  1    0.2878
## vegetation_type     3.4255  2    0.1804
## milpa               1.6884  1    0.1938
## latitude            0.0056  1    0.9406
## longitude           1.4938  1    0.2216
## harvested:vegetation_type 3.0312  2    0.2197
## harvested:milpa      0.0414  1    0.8388

```