Duff and litter: Pre- and post-thinning

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Overview

The aim of this research was to understand the impact of fuel load management (i.e., thinning) on ground fuels. Our objective was to evaluate the short-term impact of thinning on levels of duff and litter. To meet this objective, we used a "treatment and control" approach to compare fuel levels before (control, pre-thinning) and after (treatment, post-thinning) thinning was applied at five vegetation monitoring plots.

The control and treatment data sets were collected at two time points in 2021. Control data were collected before thinning, between 6/9/21 and 7/7/21. Treatment data were collected after thinning on 10/18/21 and 10/19/21, with the exception of measurements for plot FOR10 which surveyed on 12/9/21. On average, 133.6 days elapsed between control and treatment surveys (range: 104 to 183, standard deviation = 27.2).

The data for each time point were collected by conducting surveys along three transects within each of the five plots. Each transect had one quadrat for duff and litter (n = 3 quadrats per plot) and two quadrats for duff and litter (n = 6 quadrats per plot). Each time point had 30 data points each for duff and litter (5 plots x 3 transects x 2 quadrats).

We used these data to answer the following questions:

1. Did the total fuel load (all duff and litter combined) differ between control and treatment surveys?
2. Did fuel load for duff and/or litter differ between control and treatment surveys?

Fuel data for duff/litter We compared the plot-level values between control and treatment surveys (i.e., pre- and post-thinning). We defined "total (or all) duff and litter" as the combined total of the duff and litter fuel classes within a plot.

Recall that surveys were conducted along three transects within each plot; each transect had two quadrats for duff and litter (n = 6 quadrats per plot). We calculated the ***plot-level mean for each fuel class*** to account for replicate surveys within each plot. To calculated the the ***plot-level total*** at each time point, we summed the plot means for all fuel classes by fuel type.

Below is a table that shows the plot-level mean at each time point for litter and duff (total and mean by fuel class).

```
## # A tibble: 15 x 7
##    fuel_type     fuel_class plot_id `pre-thinning` post-thinnin~1 metric si_un~2
##    <fct>         <fct>      <fct>            <dbl>          <dbl> <fct>  <fct>
##  1 Duff & litter All        FOR05            6              5.33  total  Depth ~
##  2 Duff & litter All        FOR06            4.17           2.83  total  Depth ~
##  3 Duff & litter All        FOR07            5.83           3.58  total  Depth ~
```

```
##  4 Duff & litter All        FOR08           4.5           5.17  total  Depth ~
##  5 Duff & litter All        FOR10           4             1.67  total  Depth ~
##  6 Duff & litter Duff       FOR05           1             1.17  mean   Depth ~
##  7 Duff & litter Duff       FOR06           0             0.333 mean   Depth ~
##  8 Duff & litter Duff       FOR07           1             0.75  mean   Depth ~
##  9 Duff & litter Duff       FOR08           1.33          2.33  mean   Depth ~
## 10 Duff & litter Duff       FOR10           0.167         0.333 mean   Depth ~
## 11 Duff & litter Litter     FOR05           5             4.17  mean   Depth ~
## 12 Duff & litter Litter     FOR06           4.17          2.5   mean   Depth ~
## 13 Duff & litter Litter     FOR07           4.83          2.83  mean   Depth ~
## 14 Duff & litter Litter     FOR08           3.17          2.83  mean   Depth ~
## 15 Duff & litter Litter     FOR10           3.83          1.33  mean   Depth ~
## # ... with abbreviated variable names 1: `post-thinning`, 2: si_units
```

Statistical methods Statistically significant differences were identified using repeated measures analysis of variance (ANOVA) and post hoc comparisons. P-values were adjusted using the Bonferroni multiple testing correction method.

[NOTE: A repeated measures ANOVA is appropriate to use for a paired time series (before vs. after, treatment vs. control). For these data in particular, the approach accounts for elements that structure the data set in addition to time. For example, transects within each plot are more likely to be similar than transects between plots.]

The repeated-measures ANOVA is used for analyzing data where same subjects are measured more than once. This test is also referred to as a within-subjects ANOVA or ANOVA with repeated measures. The "within-subjects" term means that the same individuals (here, individuals are plots) are measured on the same outcome variable under different time points. The main goal of a repeated measures ANOVA is to evaluate if there is a statistically significant interaction effect between within-subjects factors in explaining a continuous outcome variable. The repeated measures ANOVA makes the following assumptions about the data:

- No significant outliers in any cell of the design
- Normality: the outcome (or dependent) variable should be approximately normally distributed in each cell of the design
- Assumption of sphericity: the variance of the differences between groups should be equal

We assessed outliers using the the interquartile range (IQR; IQR = Q3 - Q1). Values above Q3 + 1.5xIQR or below Q1 - 1.5xIQR are considered as outliers. Values above Q3 + 3xIQR or below Q1 - 3xIQR are considered as extreme points (or extreme outliers). Q1 and Q3 are the first and third quartile, respectively. Extreme outliers can be due to data entry errors, measurement errors, or unusual values. The outlier may be included if one believes the result will not be substantially affected; this can be evaluated by comparing the result of the ANOVA with and without the outlier.

We assessed normality by visual inspection of a QQ plot for each time point. A QQ plot draws the correlation between a given data and the normal distribution. We also conducted the Shapiro-Wilk test for each time point. Using this method, normally distributed data will have p-value >0.05.

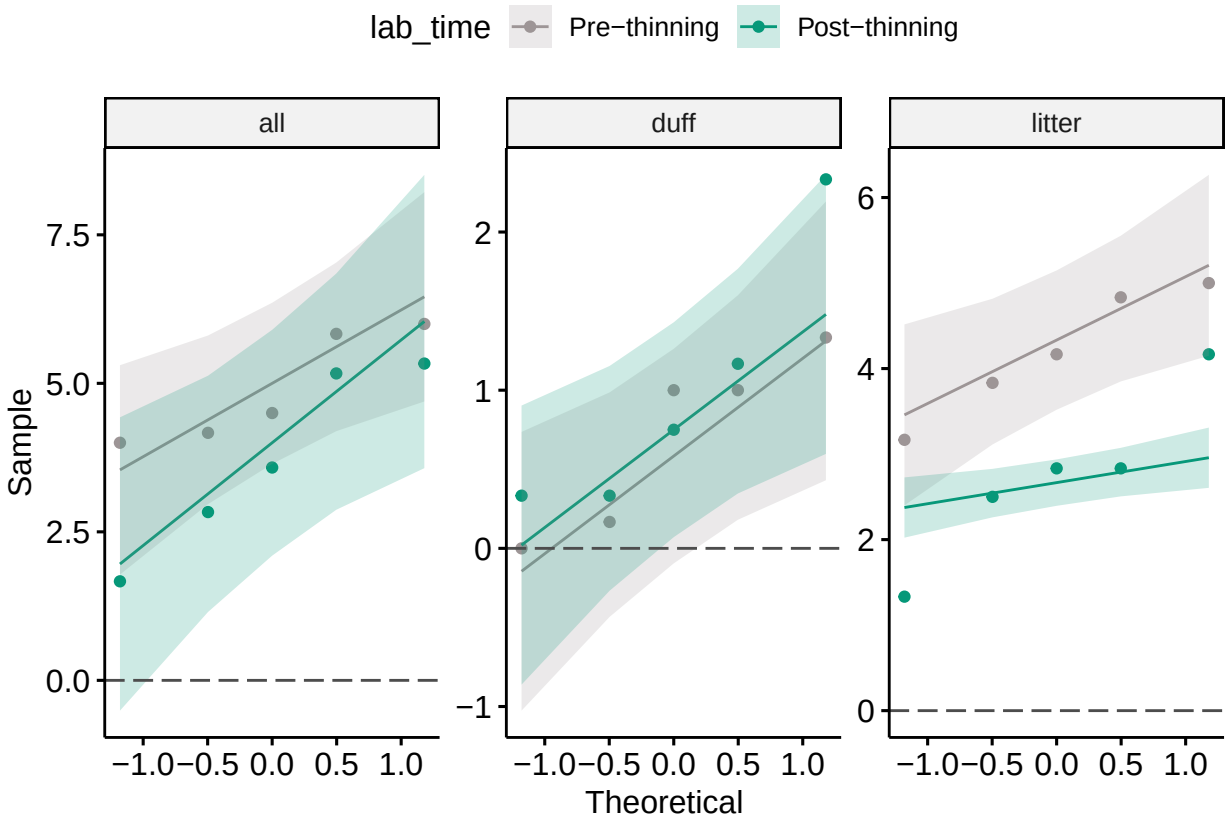
The assumption of sphericity was checked during the computation of the ANOVA test using the R function anova_test() [rstatix package]. The Mauchly's test was internally used to assess the sphericity assumption, and the Greenhouse-Geisser sphericity correction was automatically applied to factors violating the sphericity assumption.

Check assumptions

Check for outliers There were no outliers in the data set for total duff and litter. There were two extreme outliers in the post-thinning survey for litter (FOR05, FOR10).

```
## # A tibble: 2 x 8
##   fuel_type metric fuel_class time  plot_id si_value is_outlier is_extreme
##   <fct>     <fct>  <fct>      <fct> <fct>      <dbl> <lgl>      <lgl>
## 1 dl        mean   litter     t2    FOR05       4.17 TRUE       TRUE
## 2 dl        mean   litter     t2    FOR10       1.33 TRUE       TRUE
```

All points in a QQ plot created using raw values fell within the reference range for total, as well as mean duff. As expected, two outlier values for litter in the post-thinning subset strayed from the reference line (green).



Check for normality Plot-level values for total and mean duff and litter were normally distributed, as assessed by Shapiro-Wilk's test.

```
## # A tibble: 0 x 5
## # ... with 5 variables: fuel_class <fct>, time <fct>, is_normal <lgl>, p <dbl>,
## #   statistic <dbl>
```

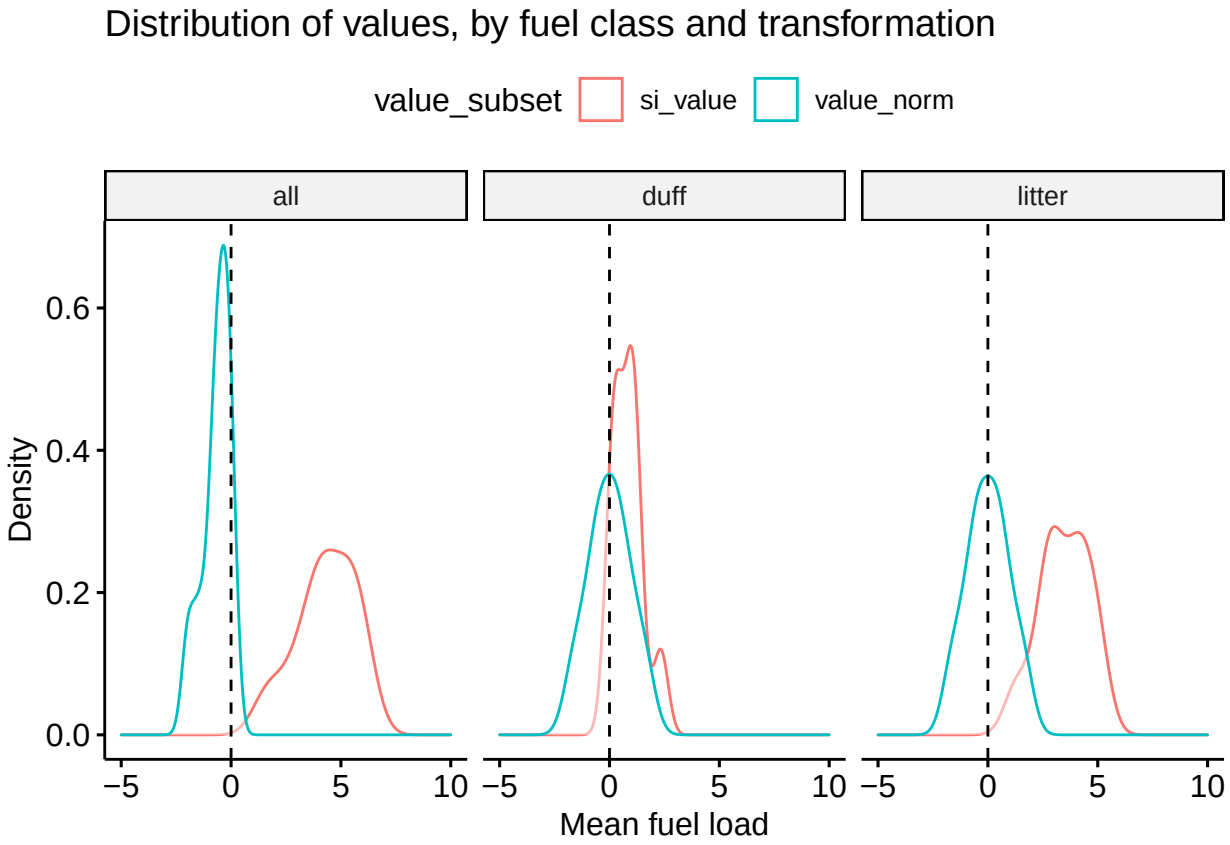
Normalize and standardize data We transformed and standardized the plot-level values for total and mean duff and litter by time point. We took this step to enable subsequent comparisons with plot-level means (which did require transformation); total values did not violate any statistical assumptions.

We subset the plot-level values for duff and litter by treatment (pre- and post-thinning), then applied an ordered quantile transformation to normalize the plot-level values for total duff and litter, calculated as:

g(x) = psiˆ-1 * ((rank(x) - .5) / (length(x)))

Where psi refers to the standard normal cumulative distribution function, rank(x) refers to each observation's rank, and length(x) refers to the number of observations. The ordered quantile transformation is a rank-based procedure by which the values of a vector are mapped to their percentile, which is then mapped to the same percentile of the normal distribution. Without the presence of ties, this essentially guarantees that the transformation leads to a uniform distribution. Values were standardized by fuel class upon normalization to have a mean of 0 and standard deviation of 1.

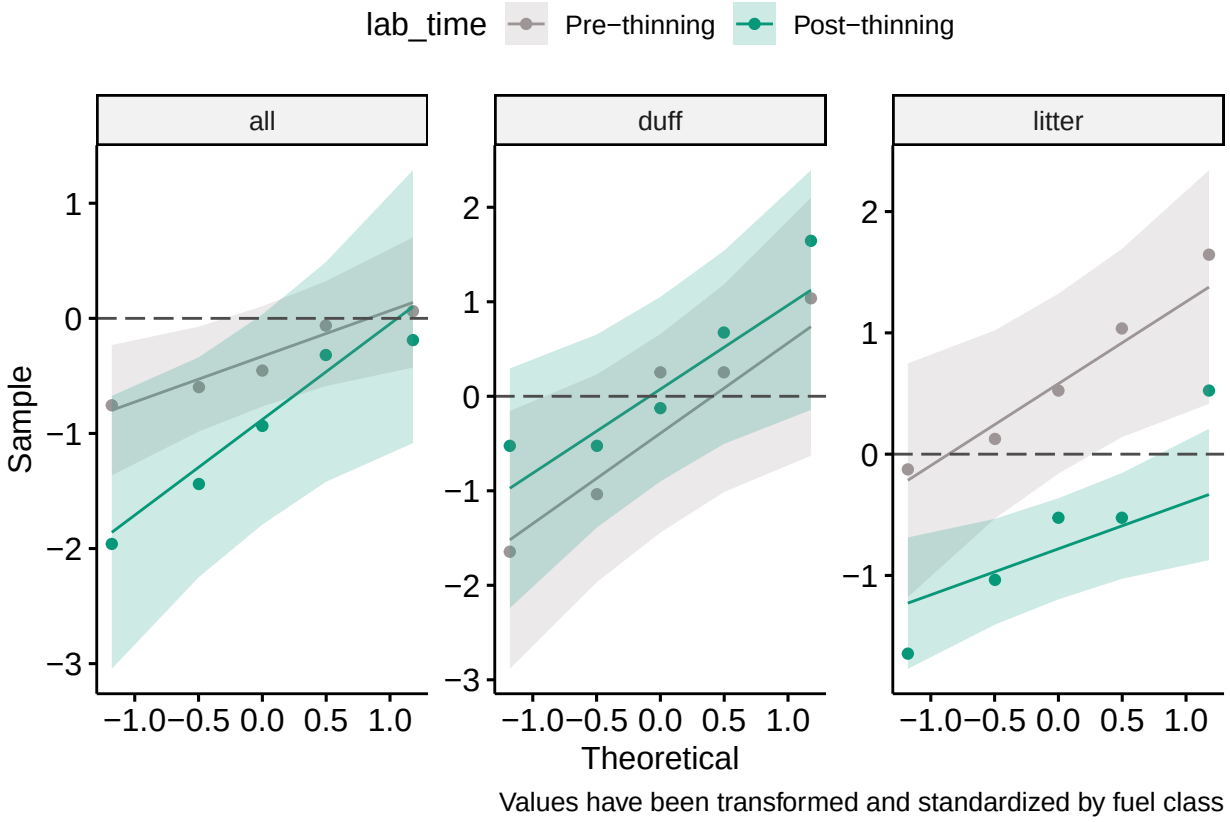
The following plot shows the raw values in pink (si_value) and the transformed values in blue (value_norm).



The transformed values for total and mean duff and litter were normally distributed, as assessed by Shapiro-Wilk's test.

```
## # A tibble: 6 x 6
##   fuel_class survey is_normal metric     p statistic
##   <fct>      <fct>  <lgl>     <fct>  <dbl>     <dbl>
## 1 all        t1     TRUE      total  0.572     0.926
## 2 all        t2     TRUE      total  0.646     0.937
## 3 duff       t1     TRUE      mean   0.582     0.928
## 4 duff       t2     TRUE      mean   0.247     0.865
## 5 litter     t1     TRUE      mean   0.810     0.96
## 6 litter     t2     TRUE      mean   0.816     0.961
```

After normalization, the outlier points on the QQ plot for litter (post-thinning, green) are nearer to the reference line, suggesting an approximately normal distribution.



Q1: Did total duff and litter differ between control and treatment surveys?

The first question we asked was whether there was a significant difference in total fuel load between surveys conduted before and after thinning treatment. We used a paired t-test to test for a significant difference between total fuel values from before and after thinning.

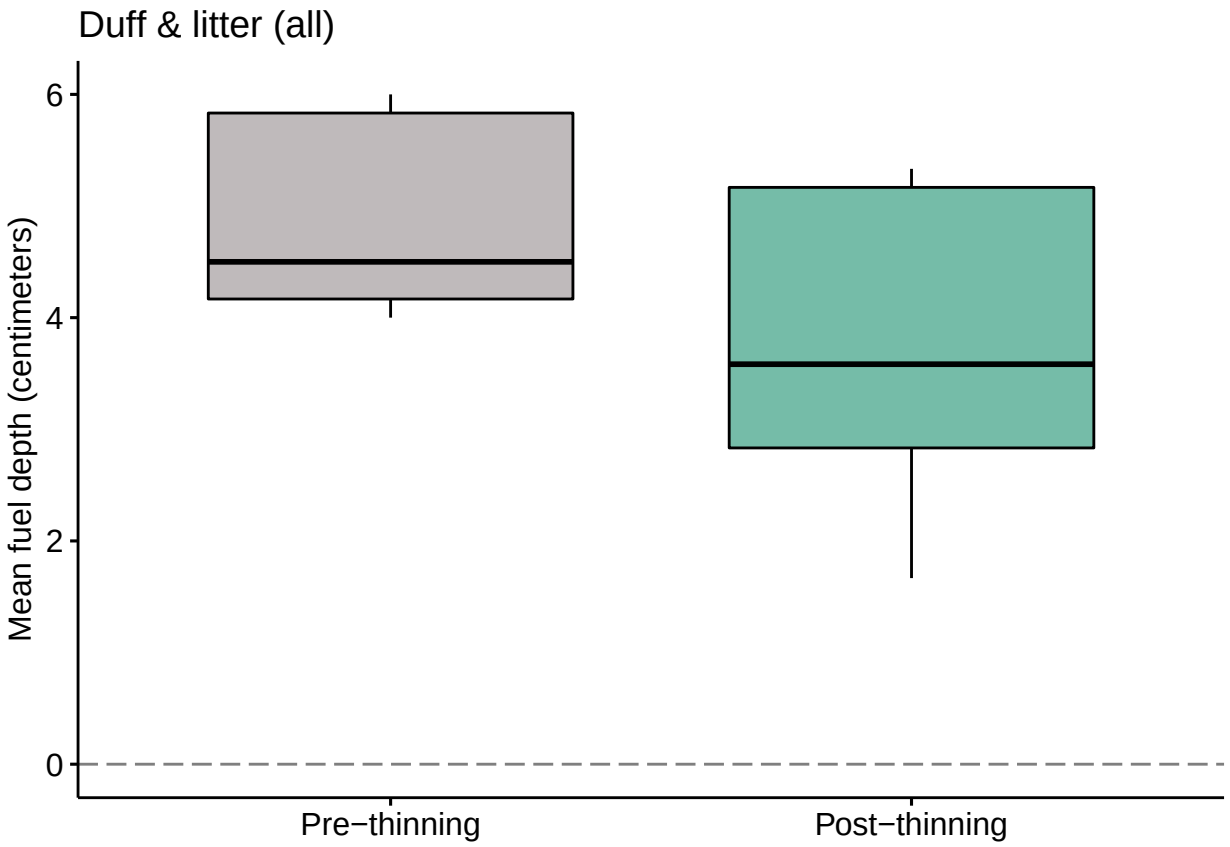
Summary statistics for the plot-level total amount of duff and litter by treatment

The following table summarizes the plot-level total amount of duff and litter for each time point.

```
## # A tibble: 2 x 7
##   fuel_type     fuel_class lab_time      si_units              mean    sd     n
##   <chr>         <chr>      <fct>         <fct>                <dbl> <dbl> <dbl>
## 1 Duff & litter All        Pre-thinning  Depth in centimeters  4.9  0.947     5
## 2 Duff & litter All        Post-thinning Depth in centimeters  3.72 1.56      5
```

Visualization of the plot-level total amount of duff and litter by treatment

A boxplot of total duff and litter by treatment showed a post-thinning decrease in depth.



Paired t-test

We found no significant difference in total fuels between pre-thin and post-thin surveys.

```
## # A tibble: 1 x 10
##   fuel_type     fuel_~1 method group1 group2 p_adj p_adj~2 stati~3    df index~4
##   <chr>         <chr>   <chr>  <chr>  <chr>  <dbl> <chr>     <dbl> <dbl> <chr>
## 1 Duff & litter All     pairw~ Pre-t~ Post-~ 0.065 n.s.       2.52     4 value_~
## # ... with abbreviated variable names 1: fuel_class, 2: p_adj_sig,
## #   3: statistic, 4: index_value
```

Q2: Did duff and litter for control and treatment surveys differ by fuel class?

Next, we investigated whether there was a significant change in plot-level fuel load by treatment when accounting for fuel class. A two-way repeated measures ANOVA was used to determine whether there was a significant interaction between treatment and fuel class on fuel load.

Here, the effect of treatment on fuel load was our focal variable of primary concern. However, the effect of treatment may differ between fuel classes, so fuel_class was considered a moderator variable.

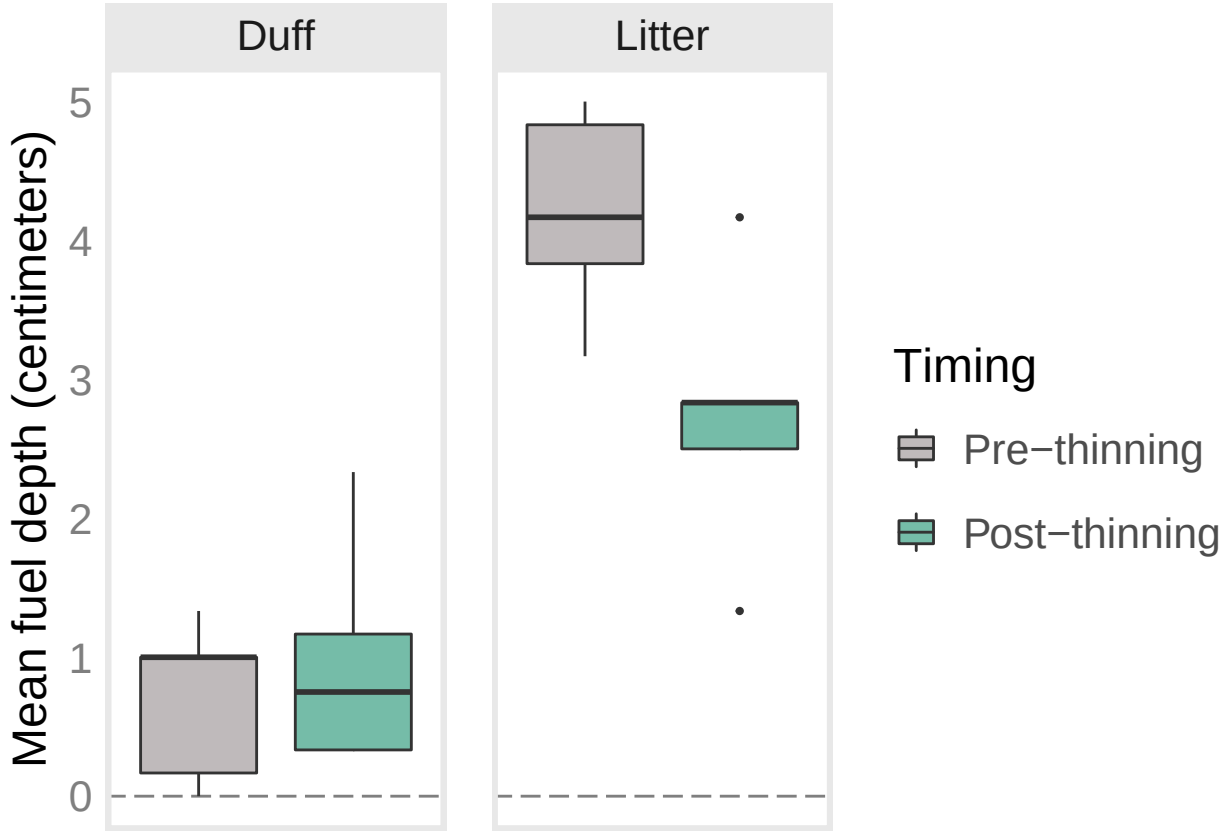
Summary statistics for duff and litter by treatment and fuel class

The following table shows the plot-level mean values for the two fuel classes (duff, litter) at each time point.

```
## # A tibble: 4 x 7
##   fuel_type     fuel_class lab_time      si_units              mean    sd     n
##   <chr>         <fct>      <fct>         <fct>                <dbl> <dbl> <dbl>
## 1 Duff & litter duff       Pre-thinning  Depth in centimeters 0.7   0.582     5
## 2 Duff & litter duff       Post-thinning Depth in centimeters 0.983 0.83      5
## 3 Duff & litter litter     Pre-thinning  Depth in centimeters 4.2   0.749     5
## 4 Duff & litter litter     Post-thinning Depth in centimeters 2.73  1.01      5
```

Visualization of duff and litter fuel classes, by treatment

We observed a decrease in mean fuel load after thinning for both fuel classes, although litter depth showed the greatest decrease.



Interaction between treatment and fuel class

There was a statistically significant two-way interaction between fuel class and treatment, $F(1, 4) = 29.238$, p-adj. $= 0.018$, ges $= 0.228$. A significant two-way interaction indicates that the impact of fuel class on fuel load depends on treatment (and vice versa).

```
## # A tibble: 1 x 11
```

```
##   fuel_t~1 fuel_~2 method effect p_adj p_adj~3 stati~4  d_fn  d_fd   ges index~5
##   <chr>    <chr>   <chr>  <chr>  <dbl> <chr>     <dbl> <dbl> <dbl> <dbl> <chr>
## 1 Duff & ~ By cla~ inter~ time:~ 0.018 *          29.2     1     4 0.228 value_~
## # ... with abbreviated variable names 1: fuel_type, 2: fuel_class,
## #   3: p_adj_sig, 4: statistic, 5: index_value
```

Effect of treatment on each fuel class

Because we found a significant interaction between fuel class and treatment, we conducted a main effect assessment for each fuel class. We found a significant main effect of treatment on fuel load for the litter fuel class (p-adj. = 0.012), but not for duff.

```
## # A tibble: 2 x 11
##   fuel_t~1 fuel_~2 method effect p_adj p_adj~3 stati~4  d_fn  d_fd   ges index~5
##   <chr>    <fct>   <chr>  <chr>  <dbl> <chr>     <dbl> <dbl> <dbl> <dbl> <chr>
## 1 Duff & ~ litter  main ~ time   0.012 *         27.4      1     4 0.473 value_~
## 2 Duff & ~ duff    main ~ time   0.264 n.s.       3.58     1     4 0.06  value_~
## # ... with abbreviated variable names 1: fuel_type, 2: fuel_class,
## #   3: p_adj_sig, 4: statistic, 5: index_value
```

A post hoc pairwise comparison showed a significant difference in fuel load between treatments for litter. No other comparisons were significant.

```
## # A tibble: 2 x 10
##   fuel_type     fuel_~1 method group1 group2 p_adj p_adj~2 stati~3    df index~4
##   <chr>         <fct>   <chr>  <chr>  <chr>  <dbl> <chr>     <dbl> <dbl> <chr>
## 1 Duff & litter litter  pairw~ t1     t2     0.006 **         5.23     4 value_~
## 2 Duff & litter duff    pairw~ t1     t2     0.132 n.s.      -1.89     4 value_~
## # ... with abbreviated variable names 1: fuel_class, 2: p_adj_sig,
## #   3: statistic, 4: index_value
```
