

Duff and litter: Pre- and post-Tubbs fire

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Overview

The aim of this research was to understand the impact of the 2017 Tubbs fire on fuel loads. The objective of this research was to evaluate the impact of the 2017 Tubbs fire on duff and litter fuel loads over time. To meet this objective, we compared the pre- and post-fire levels of duff and litter at nine long-term vegetation monitoring plots. Surveys were conducted at four time points: 2016 (1y pre-fire), 2017 (0y post-fire), 2019 (2y post-fire), and 2021 (4y post-fire). We conducted surveys along two transects per plot; each transect had two quadrats. As a result, each time point had 36 data points: 9 plots x 2 transects x 2 quadrats.

Using the duff and litter data set as input, we asked the following questions:

1. Did the total fuel load (all duff and litter combined) differ between years?
2. Did fuel load differ between years for duff and/or litter? If so, which years were different?

Fuel measurements for duff and litter depth We answered these questions using a time series of plot-level data for (1) the total amount of duff and litter and (2) the mean amount of each fuel class (duff, litter). We defined “total (or all) duff and litter” as the combined total of the duff and litter within a plot.

We calculated the *plot-level mean for each fuel class* to account for replicate surveys within each plot. To calculate 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).

fuel_class	plot_id	metric	units	2016	2017	2019	2021
All	RxF01	total	cm	4.445	1.714	3.842	4.540
All	RxF02	total	cm	10.001	2.445	8.477	1.841
All	RxF03	total	cm	7.620	4.413	1.397	1.143
All	RxF04	total	cm	6.668	1.746	3.937	3.651
All	RxF05	total	cm	4.127	0.953	6.096	1.016
All	RxF06	total	cm	7.652	1.302	9.970	1.365
All	RxF07	total	cm	11.748	1.810	4.318	3.111
All	RxF08	total	cm	8.731	1.048	2.699	3.080
All	RxF09	total	cm	10.160	1.238	1.810	1.429
Duff	RxF01	mean	cm	2.857	0.794	0.603	0.254
Duff	RxF02	mean	cm	3.969	1.270	0.064	0.254
Duff	RxF03	mean	cm	2.699	1.746	0.127	0.286
Duff	RxF04	mean	cm	3.651	1.238	0.286	1.270
Duff	RxF05	mean	cm	2.540	0.603	0.064	0.254
Duff	RxF06	mean	cm	5.239	1.111	0.444	0.254
Duff	RxF07	mean	cm	6.033	1.270	0.349	0.254
Duff	RxF08	mean	cm	5.080	0.857	0.000	0.254
Duff	RxF09	mean	cm	3.810	1.111	0.000	0.254
Litter	RxF01	mean	cm	1.587	0.921	3.239	4.286
Litter	RxF02	mean	cm	6.032	1.175	8.414	1.587
Litter	RxF03	mean	cm	4.921	2.667	1.270	0.857
Litter	RxF04	mean	cm	3.016	0.508	3.651	2.381
Litter	RxF05	mean	cm	1.587	0.349	6.032	0.762
Litter	RxF06	mean	cm	2.413	0.190	9.525	1.111
Litter	RxF07	mean	cm	5.715	0.540	3.969	2.857
Litter	RxF08	mean	cm	3.651	0.190	2.699	2.826
Litter	RxF09	mean	cm	6.350	0.127	1.810	1.175

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.

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 $Q3 + 1.5 \times IQR$ or below $Q1 - 1.5 \times IQR$ are considered as outliers. Values above $Q3 + 3 \times IQR$ or below $Q1 - 3 \times IQR$ 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.

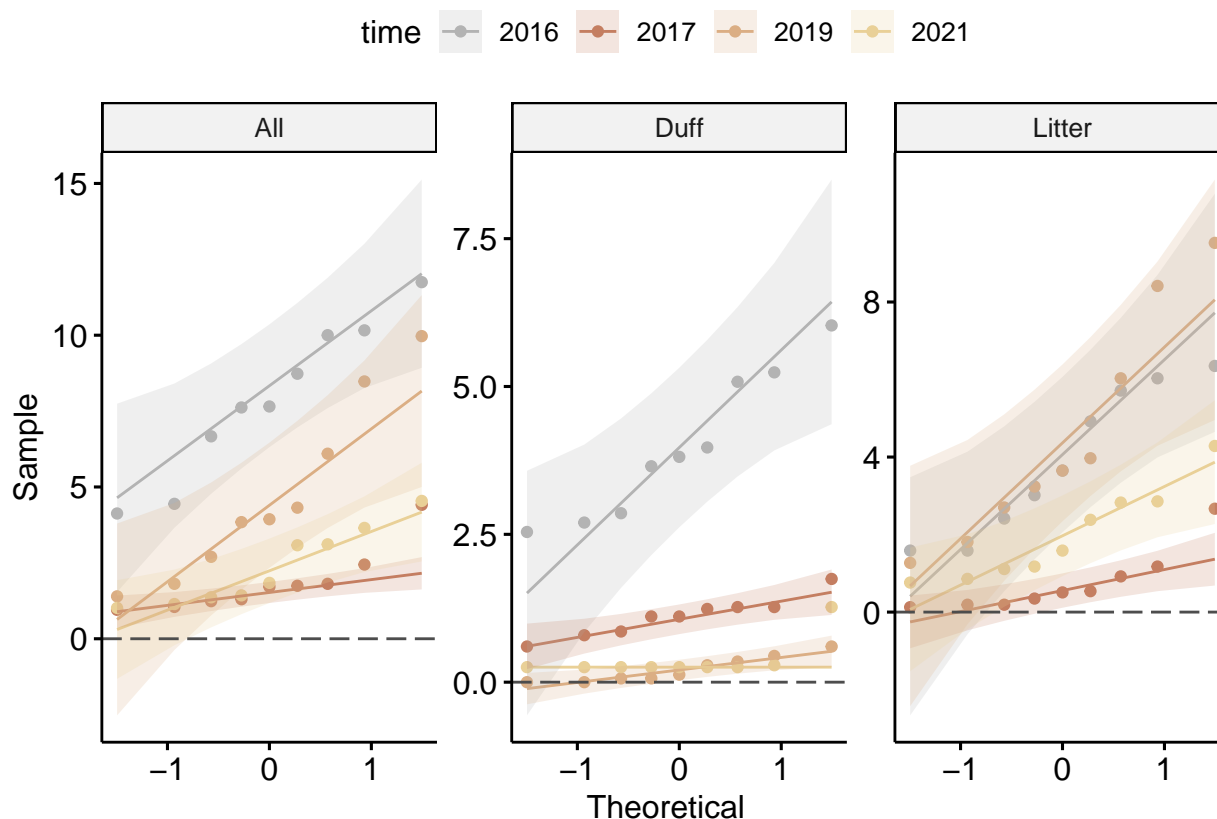
The assumption of sphericity was checked during the computation of the ANOVA test using the R function `anova_test()` [rstatix package]. The Mauchly's test was internally used to assess the sphericity assumption, and the Greenhouse-Geisser sphericity correction was automatically applied to factors violating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

Check assumptions

Check for outliers There was one extreme outlier in the data set for total duff and litter (plot RxF03 in 2017). There were two extreme outliers in the mean by fuel class data set for duff (plots RxF03 and RxF04 in 2021).

fuel_class	metric	time	plot_id	si_value	units	is_outlier	is_extreme
all	total	2017	RxF03	4.413	cm	TRUE	TRUE
duff	mean	2021	RxF03	0.286	cm	TRUE	TRUE
duff	mean	2021	RxF04	1.270	cm	TRUE	TRUE

A QQ plot created using raw values showed one point for total (all) duff and litter in 2017 (red) that was distant from the reference line. As expected, two outlier values for duff in 2021 (yellow) strayed from the reference line.



Check for normality Total duff and litter load was normally distributed at each time point ($p > 0.01$), except for year 2017. Plot-level values by fuel class were normally distributed at each time point ($p > 0.01$),

except for duff in 2021 and litter in 2017. Normality was assessed by Shapiro-Wilk's test.

fuel_class	time	is_normal	p	statistic
all	2017	FALSE	0.0094872	0.771
litter	2017	FALSE	0.0066324	0.758
duff	2021	FALSE	0.0000005	0.406

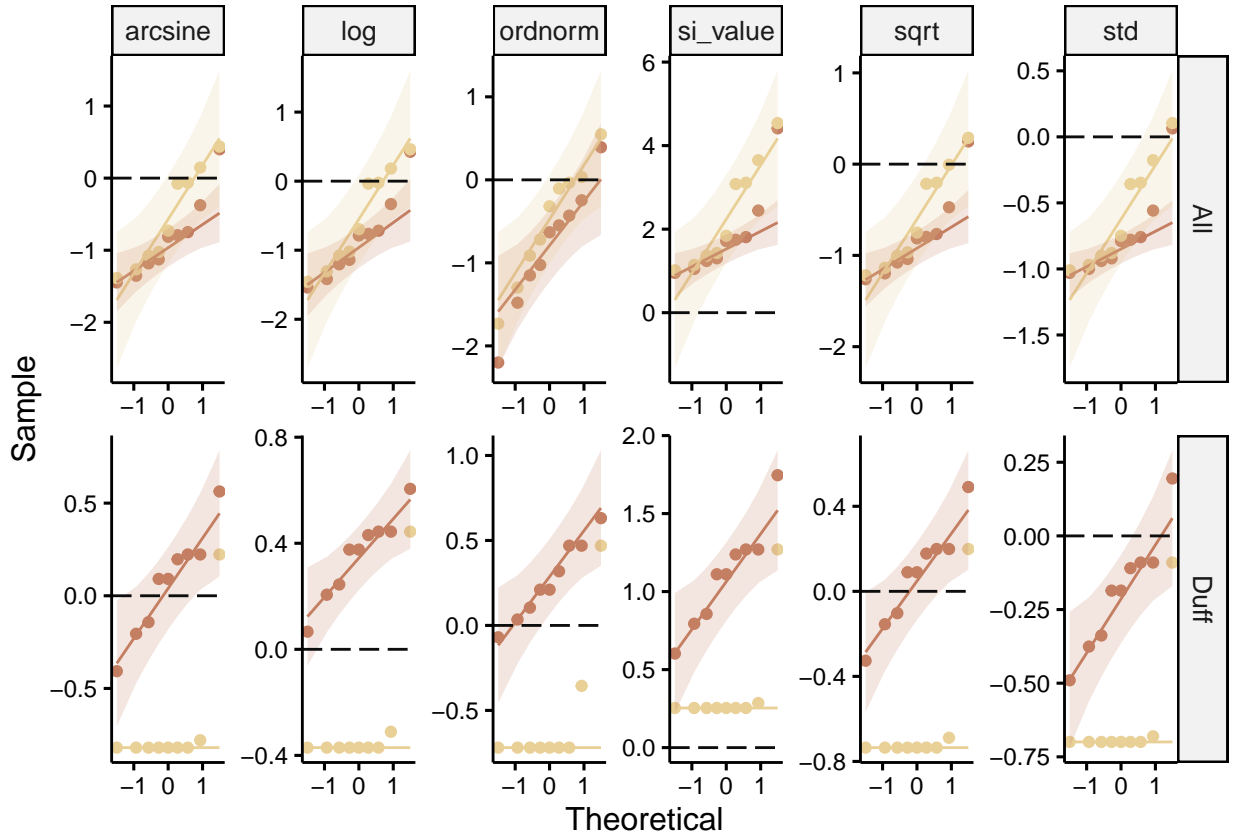
Identify the most appropriate transformation We applied a series of transformations to each subset, then evaluated the resulting data for outliers and normality. To create the transformed subsets, we used `scale()` from base R and four normalization functions from the `bestNormalize` package: `arcsinh_x`, `log_x`, `orderNorm`, and `sqrt_x`.

The application of the arcsine and log functions resolved outliers within the subset for total fuels. Outliers within the Duff subset were not improved with any transformation.

fuel_class	time	method	plot_id	number	is_outlier	is_extreme
all	2017	value_arcsine	RxF03	0.3998317	TRUE	FALSE
all	2017	value_log	RxF03	0.4250334	TRUE	FALSE

A QQ plot of the transformed values for fuel classes with outliers (rows; all fuels, duff) by transformation (columns) is below.

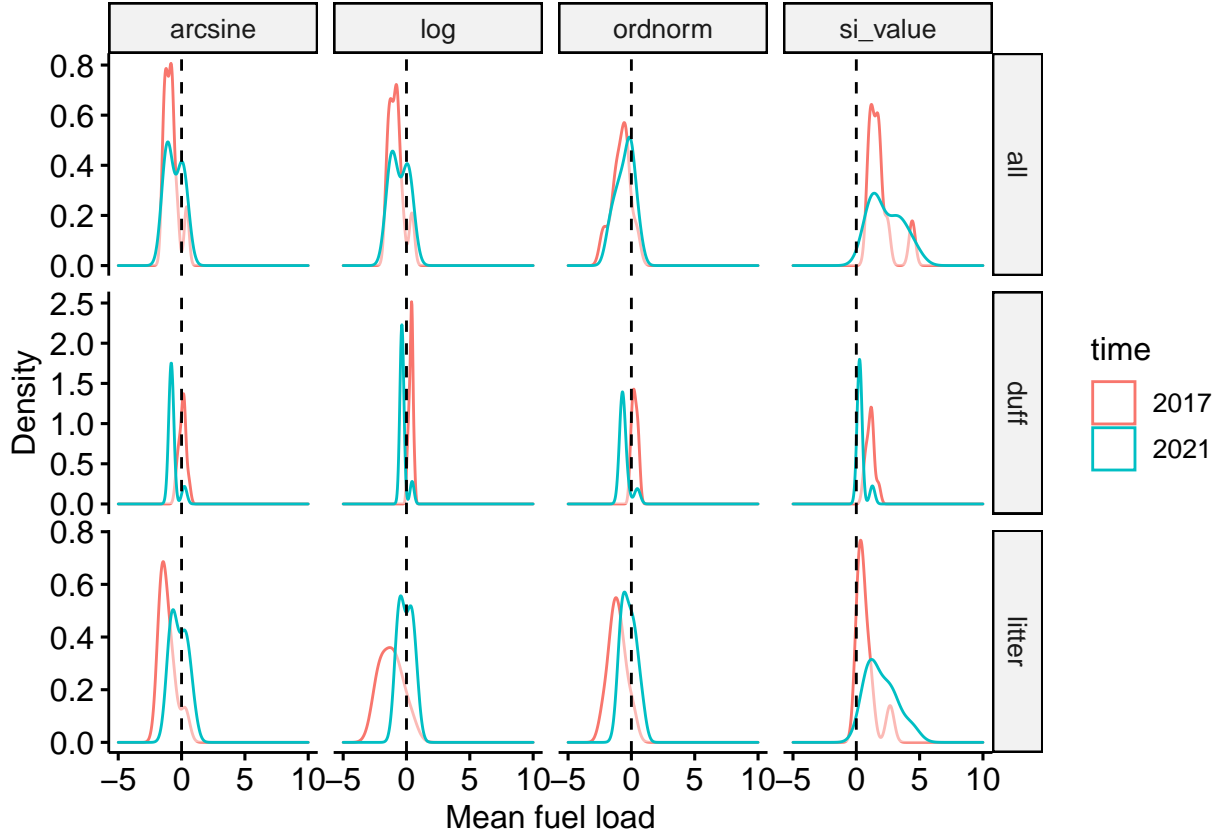
The outlier points for total fuels in 2017 (yellow) are nearer to the reference line for the arcsine, log, and `orderNorm` transformations.



The application of the multiple transformations resulted in normally-distributed values for total fuels and litter in 2017; raw data for these two subsets were not normally distributed. The effective transformations for total fuels and litter included: arcsine, log, `orderNorm`, and square root. No transformation normalized the distribution of the Duff subset in 2021.

fuel_class	time	value_arcsine	value_log	value_norm	value_ordnorm	value_sqrt
all	2017	TRUE	TRUE	TRUE	TRUE	TRUE
litter	2017	TRUE	TRUE	TRUE	TRUE	TRUE

The following plot shows the values by fuel class (rows) and transformation method (columns). Overall, the orderNorm function (green) appeared to be an effective normalization method for the total and litter subsets.



Normalize and standardize data We transformed and standardized the plot-level values for total fuels and mean by fuel class by time point. We subset the plot-level values by fuel class (all, duff, litter), then applied an ordered quantile transformation to normalize the values, calculated as:

$$g(x) = \psi^{-1} * ((\text{rank}(x) - .5) / (\text{length}(x)))$$

Where ψ refers to the standard normal cumulative distribution function, $\text{rank}(x)$ refers to each observation's rank, and $\text{length}(x)$ 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.

Total fuel load

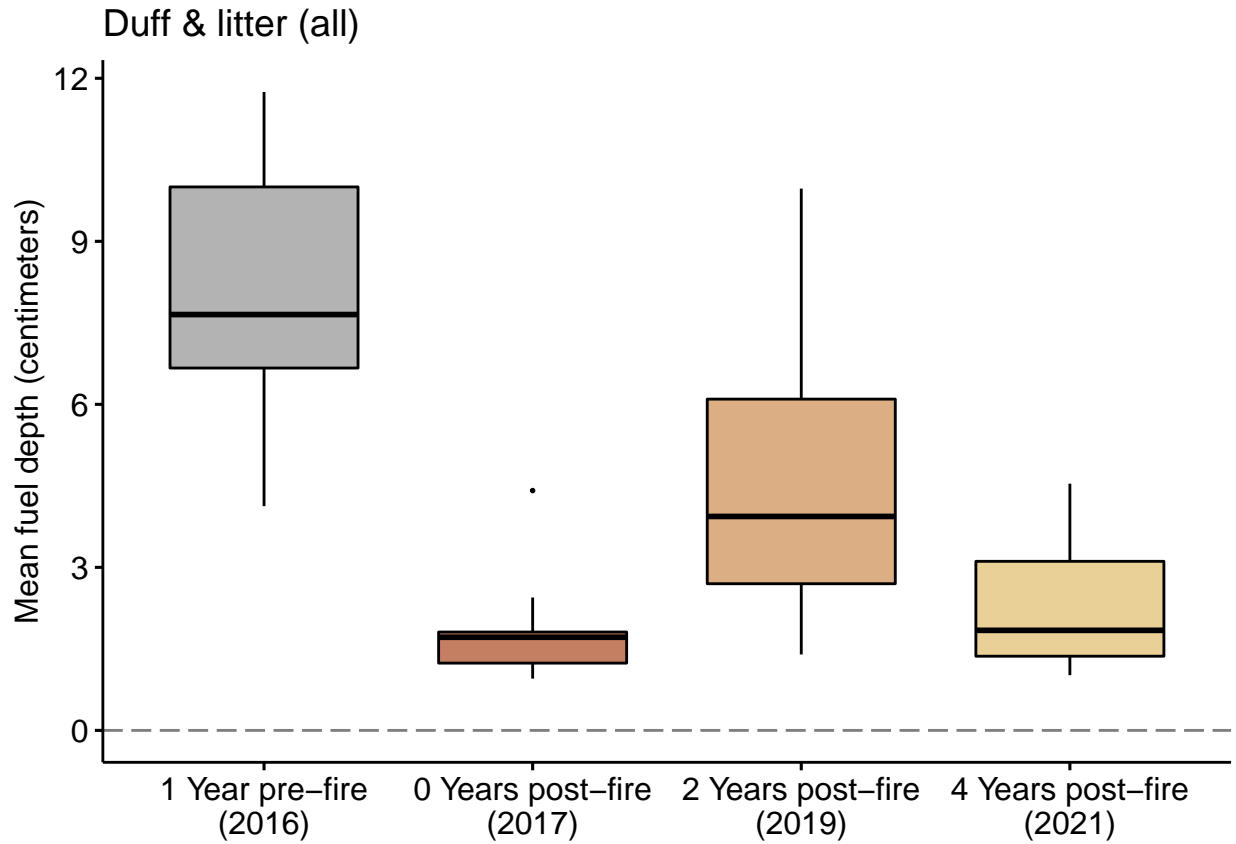
Summary statistics for total fuel load by year

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

fuel_class	time	units	mean	sd	n
All	2016	cm	7.906	2.570	9
All	2017	cm	1.852	1.065	9
All	2019	cm	4.727	2.931	9
All	2021	cm	2.353	1.271	9

Visualization of total fuel load by year

The visual inspection of a box plot of total duff and litter by year showed lower values in post-fire years. Specifically, we observed a decrease in the depth of all duff and litter immediately following the fire (2017; 0-y post-fire). The mean depth of duff and litter did not return to pre-fire levels in subsequent years.



Main effect of year on total fuel load

We conducted a one-way repeated measures ANOVA to test for a main effect of year on fuel load (i.e., whether total fuel load was significantly different between the four time points). We found a significant effect of year on total fuel load $F(3, 24) = 14.413$, $p\text{-adj.} = 1.41e-05$, $ges = 0.554$.

fuel_class	method	effect	p_adj	p_adj_sig	statistic	d_fn	d_fd	ges	index_value
All	me	time	1.41e-05	***	14.413	3	24	0.554	value_norm

Pairwise comparison of total fuel load between years

Because we found a significant effect of year, we conducted a post hoc pairwise comparison in total fuel load between years. We found a significant difference in total fuel load between 2016 and 2017 (p-adj. <0.001) and 2021 (p-adj <0.01).

fuel_class	method	group1	group2	p_adj	p_adj_sig	statistic	df	index_value
All	pwc	2016	2017	0.000834	***	6.793	8	value_norm
All	pwc	2016	2021	0.002000	**	5.787	8	value_norm
All	pwc	2017	2019	0.147000	n.s.	-2.764	8	value_norm
All	pwc	2016	2019	0.169000	n.s.	2.676	8	value_norm
All	pwc	2019	2021	0.341000	n.s.	2.224	8	value_norm
All	pwc	2017	2021	1.000000	n.s.	-1.054	8	value_norm

Mean fuel load, by fuel class

Summary statistics for mean fuel load by fuel class and year

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

fuel_class	time	units	mean	sd	n
duff	2016	cm	3.986	1.231	9
duff	2017	cm	1.111	0.334	9
duff	2019	cm	0.215	0.216	9
duff	2021	cm	0.370	0.337	9
litter	2016	cm	3.919	1.892	9
litter	2017	cm	0.741	0.803	9
litter	2019	cm	4.512	2.881	9
litter	2021	cm	1.982	1.187	9

Visualization of mean fuel load by fuel class and year

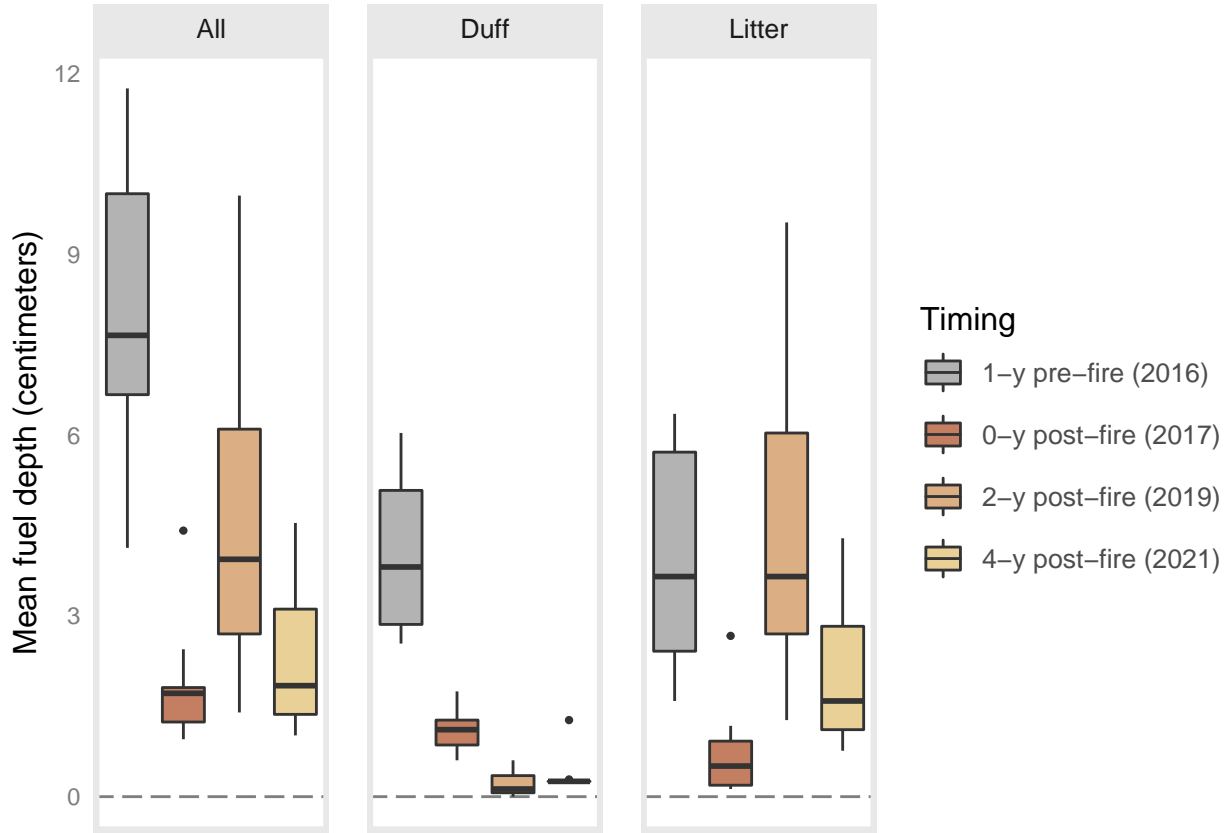
A visual inspection of box plots for mean fuel depth by year showed differing post-fire trends between the duff and litter fuel classes. Although duff and litter both showed a decrease in depth immediately following the fire (2017, 0-y post-fire; red), the subsequent annual trends differed. The temporal trend for total duff and litter was more similar to that of the litter fuel class, and showed less similarity to that of the duff fuel class. These results suggest that litter was an influential driver of the post-fire patterns observed for duff and litter as a group.

Duff

- Duff depth remained low in all post-fire years.
- Values in 2019 (orange) and 2021 (yellow) were lower than those for 2017 (red).

Litter

- Litter depth returned to pre-fire levels 2-y after the fire (2019, orange), then showed a second decrease in 2021 (4-y post-fire, yellow).
- This post-fire increase and decrease in litter depth was similar to the temporal trend observed for duff and litter overall.



Interaction between year and fuel class

We investigated whether there was a significant change in the plot-level fuel load over time when accounting for fuel class (duff, litter). A two-way repeated measures ANOVA was used to determine whether there was a significant interaction between year and fuel class on mean fuel load. Here, the effect of year on fuel load was our focal variable of primary concern. However, the effect of year may differ between duff and litter, so fuel_class was considered a moderator variable.

There was a statistically significant two-way interaction between fuel clas and year, $F(3, 24) = 26.727$, $p\text{-adj.} = 2.397\text{e-}07$, $ges = 0.506$. A significant two-way interaction indicates that the impact of fuel class on fuel load depends on year (and vice versa).

method	effect	p_adj	p_adj_sig	statistic	d_fn	d_fd	ges	index_value
interaction	time:fuel_class	2e-07	***	26.727	3	24	0.506	value_norm

Effect of year on each fuel class

Because we found a significant interaction between fuel class and year, we conducted a main effect assessment for each fuel class.

We found a significant main effect of year on fuel load for both fuel classes at a significance level of <0.001 .

fuel_class	method	effect	p_adj	p_adj_sig	statistic	d_fn	d_fd	ges	index_value
duff	me	time	0.0000000	***	39.225	3	24	0.762	value_norm
litter	me	time	0.0001138	***	11.899	3	24	0.556	value_norm

Pairwise comparisons of mean fuel load between years, by fuel class

The pre-fire fuel load (2016) was significantly different from that in 2017 for both fuel classes (duff: $p < 0.001$; litter: $p < 0.01$). For each fuel class, we found a significant difference in mean fuel load between the years for the following comparisons:

Duff

- 2016 and 2019, 2021, 2017 ($p\text{-adj} < 0.001$)
- 2017 and 2019, 2021 ($p\text{-adj} < 0.01$)
- A pairwise comparison between years showed that the mean duff depth was significantly different between all years except 2019/2021.

Litter

- 2017 and 2016, 2019 ($p\text{-adj} < 0.01$)

fuel_class	method	group1	group2	p_adj	p_adj_sig	statistic	df	index_value
duff	pwc	2016	2019	0.000155	***	8.599	8	value_norm
duff	pwc	2016	2021	0.000356	***	7.664	8	value_norm
duff	pwc	2016	2017	0.001000	***	6.294	8	value_norm
duff	pwc	2017	2021	0.002000	**	6.040	8	value_norm
duff	pwc	2017	2019	0.008000	**	4.784	8	value_norm
duff	pwc	2019	2021	0.846000	n.s.	-1.634	8	value_norm
litter	pwc	2016	2017	0.004000	**	5.387	8	value_norm
litter	pwc	2017	2019	0.008000	**	-4.847	8	value_norm
litter	pwc	2017	2021	0.085000	n.s.	-3.128	8	value_norm
litter	pwc	2016	2021	0.204000	n.s.	2.554	8	value_norm
litter	pwc	2019	2021	0.209000	n.s.	2.536	8	value_norm
litter	pwc	2016	2019	1.000000	n.s.	-0.395	8	value_norm