REX: Insect Preference Trial Plots & Analyses

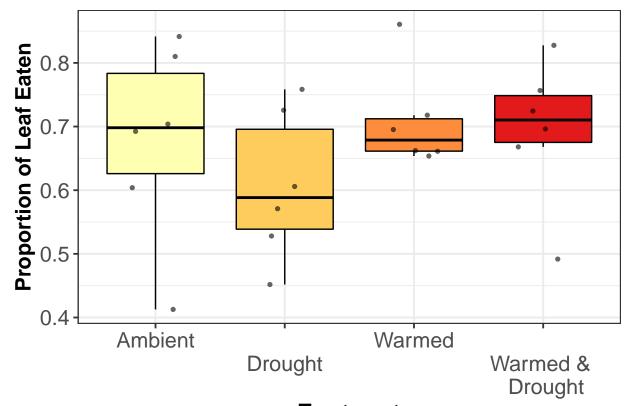
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September 17, 2021

```
## Loading required package: stats4
## Loading required package: lme4
## Loading required package: Matrix
## Attaching package: 'lmerTest'
## The following object is masked from 'package:lme4':
##
##
       lmer
## The following object is masked from 'package:stats':
##
       step
## Loading required package: MASS
## Loading required package: survival
## Learn more about sjPlot with 'browseVignettes("sjPlot")'.
## Loading required package: carData
## Registered S3 methods overwritten by 'car':
##
     method
                                      from
##
     influence.merMod
     cooks.distance.influence.merMod lme4
##
     dfbeta.influence.merMod
                                      lme4
     dfbetas.influence.merMod
                                      1me4
## -- Attaching packages ------ 1.3.0 --
## v ggplot2 3.3.3
                                  0.3.4
                      v purrr
## v tibble 3.0.6 v dplyr 1.0.4
## v tidyr 1.1.2 v stringr 1.4.0
## v readr 1.4.0 v forcats 0.5.1
```

```
## -- Conflicts ----- tidyverse_conflicts() --
## x tidyr::expand() masks Matrix::expand()
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                    masks stats::lag()
## x tidyr::pack()
                    masks Matrix::pack()
## x dplyr::recode() masks car::recode()
## x dplyr::select() masks MASS::select()
## x dplyr::slice() masks bbmle::slice()
## x purrr::some()
                   masks car::some()
## x tidyr::unpack() masks Matrix::unpack()
##
## Attaching package: 'rstatix'
## The following object is masked from 'package:MASS':
##
##
      select
## The following object is masked from 'package:stats':
##
      filter
##
## Attaching package: 'janitor'
## The following object is masked from 'package:rstatix':
##
##
      make_clean_names
## The following objects are masked from 'package:stats':
##
##
      chisq.test, fisher.test
   [1] "animal"
##
##
   [2] "GHG"
   [3] "Irrigation"
##
## [4] "lookup_tables"
## [5] "microbes"
## [6] "REX Microbial sampling IDs complete.gsheet"
## [7] "REX_Data_Management.gdoc"
## [8] "REX_Data_Template.gsheet"
## [9] "REX_warmx_metadata.csv"
## [10] "REX_warmx_metadata.xlsx"
## [11] "soil"
## [12] "T7_ANPP"
## [13] "T7_plant_comp"
## [14] "T7_plant_phenology"
## [15] "T7_warmx_insect"
## [16] "T7_warmx_plant_traits"
## [17] "T7 warmx VOC"
## [18] "Untitled form.gform"
## [19] "weather"
```

```
# Clean data
insects <- insects[-88, ] #this data point had a greater SLA_after than it's SLA_before even though it
# getting rid of NAs in data. NAs in data are from leaf
# samples collected for CN analyses and not used in the
# trials
insects1 <- na.omit(insects)</pre>
# delete date column
insects1$date <- NULL</pre>
# Change column names to lowercase
names(meta) <- tolower(names(meta))</pre>
# Data Wrangling
# Take subplot average of sla before trial
sla_before <- insects1 %>% group_by(replicate, footprint_location,
    subplot_location) %>% dplyr::summarize(sla_before_mean = mean(sla_before,
   na.rm = TRUE))
## 'summarise()' has grouped output by 'replicate', 'footprint_location'. You can override using the '.
# Take subplot average of sla after trial
sla_after <- insects1 %>% group_by(replicate, footprint_location,
    subplot_location) %>% dplyr::summarize(sla_after_mean = mean(sla_after_1,
   na.rm = TRUE))
## 'summarise()' has grouped output by 'replicate', 'footprint_location'. You can override using the '.
insects2 <- left_join(sla_before, sla_after) # combine sla_before and sla_after into one dataframe
## Joining, by = c("replicate", "footprint_location", "subplot_location")
# calculate the proportion of leaves eaten
insects3 <- insects2 %>% group_by(replicate, footprint_location,
    subplot_location) %>% dplyr::summarize(prop_eaten = sla_after_mean/sla_before_mean)
## 'summarise()' has grouped output by 'replicate', 'footprint_location'. You can override using the '.
insects4 <- left_join(insects2, insects3)</pre>
## Joining, by = c("replicate", "footprint_location", "subplot_location")
insects5 <- left_join(meta, insects4)</pre>
## Joining, by = c("replicate", "footprint_location", "subplot_location")
# get rid of NAs in data - irrigated control plots were not
# used in this experiment
insects6 <- na.omit(insects5)</pre>
```



Treatment

```
# dev.off()

# looking at the graph, there is more variability in the
# grasshoppers eating of ambient leaves and less so going
# from ambient to warmed + drought leaves.

# subset the leaves that had no herbivory - did certain
# treatments have more no herbivory on it's leaves?
no_herbivory <- subset(insects1, notes == "no_herbivory")
no_herbivory <- left_join(no_herbivory, meta)

## Joining, by = c("replicate", "footprint_location", "subplot_location")</pre>
```

```
##
    replicate
                      footprint_location subplot_location
                                                          plant_number
##
  Length:36
                            :1.000
                                        Length:36
                                                          Length:36
                     Min.
  Class : character
                     1st Qu.:2.000
                                        Class :character
                                                          Class : character
## Mode :character
                     Median :3.000
                                       Mode :character
                                                          Mode :character
##
                     Mean
                            :3.194
##
                     3rd Qu.:4.250
##
                     Max. :7.000
##
     sla_before
                    sla after
                                  sla_after_1
                                                    notes
         :1.340 Min. :1.080 Min. :1.340 Length:36
## Min.
  1st Qu.:2.390    1st Qu.:1.755    1st Qu.:2.390
                                                 Class : character
## Median :3.075 Median :2.610 Median :3.075
                                                 Mode :character
## Mean :3.255 Mean :2.801 Mean :3.255
## 3rd Qu.:3.857
                  3rd Qu.:3.458
                                  3rd Qu.:3.857
## Max. :6.000 Max. :5.670 Max. :6.000
     id_number
                                     footprint
##
                   treatment
                                                         subplot
         :185.0 Length:36
## Min.
                                    Length:36
                                                       Length:36
## 1st Qu.:238.5 Class :character Class :character
                                                       Class : character
## Median :293.0 Mode :character Mode :character
                                                       Mode :character
## Mean
         :274.1
## 3rd Qu.:321.0
## Max. :327.0
## subplot_descriptions unique_id
                       Length:36
## Length:36
## Class :character
                       Class :character
## Mode :character
                       Mode :character
##
##
##
tabyl(no_herbivory, subplot_descriptions) # pretty even among treatments
##
   subplot_descriptions n
                            percent
##
                ambient 10 0.2777778
##
                drought 10 0.2777778
##
                 warmed 8 0.222222
##
         warmed_drought 8 0.222222
# proportion of leaves eaten at individual level
indv_prop <- insects1 %>% group_by(replicate, footprint_location,
   subplot_location, plant_number) %>% dplyr::summarize(prop_eaten = sla_after_1/sla_before)
## 'summarise()' has grouped output by 'replicate', 'footprint_location', 'subplot_location'. You can o
indv_prop <- left_join(indv_prop, insects1)</pre>
## Joining, by = c("replicate", "footprint_location", "subplot_location", "plant_number")
```

summary(no_herbivory)

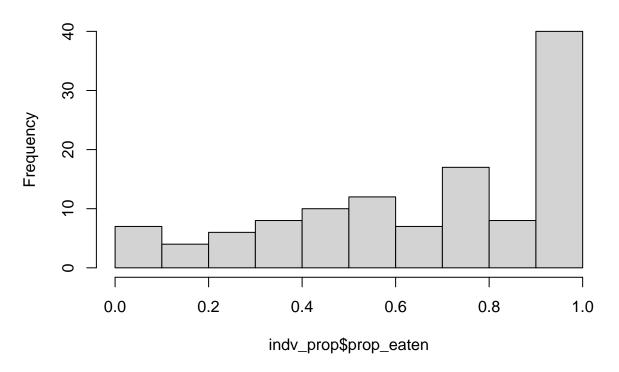
```
indv_prop <- left_join(meta, indv_prop)

## Joining, by = c("replicate", "footprint_location", "subplot_location")

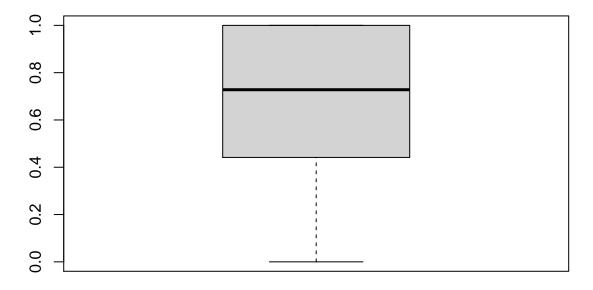
indv_prop <- na.omit(indv_prop)

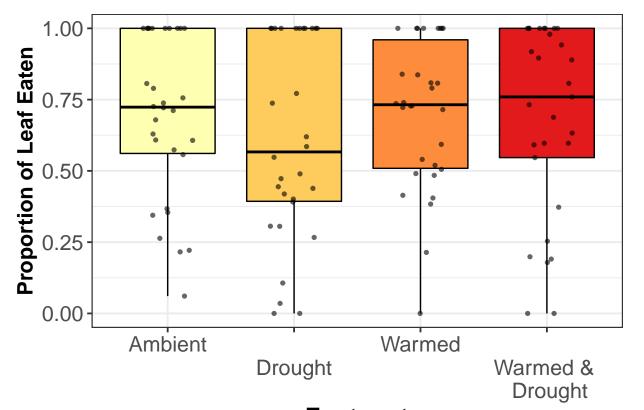
hist(indv_prop$prop_eaten)</pre>
```

Histogram of indv_prop\$prop_eaten



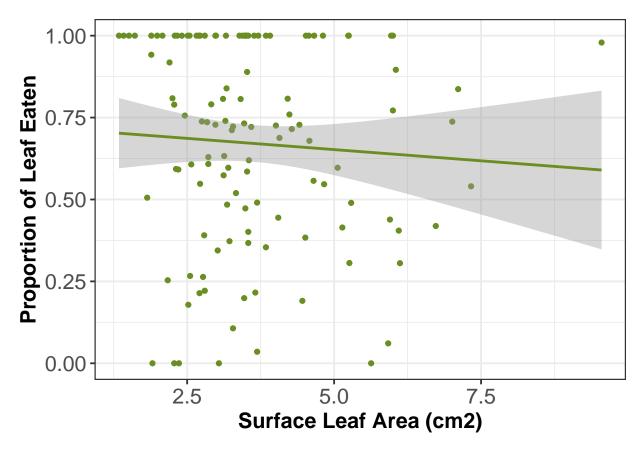
boxplot(indv_prop\$prop_eaten)





Treatment

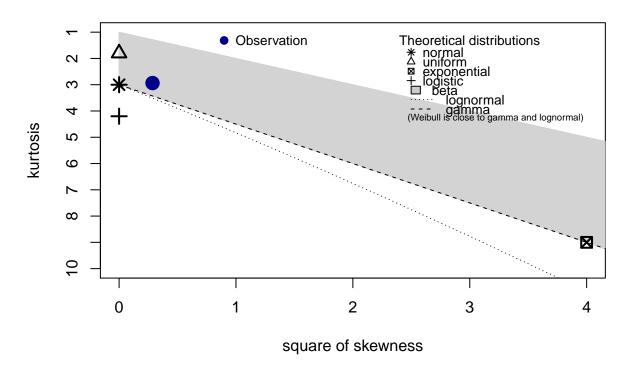
'geom_smooth()' using formula 'y ~ x'



```
# dev.off()
summary(lm_indv_prop)
```

```
##
## Call:
## lm(formula = prop_eaten ~ sla_before, data = indv_prop)
##
## Residuals:
       Min
                 1Q
                      Median
                                           Max
## -0.69457 -0.20510 0.05407 0.31091 0.38906
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 0.72071
                          0.07849
                                    9.182 1.83e-15 ***
## sla_before -0.01369
                          0.02012 -0.680
                                             0.498
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3091 on 117 degrees of freedom
## Multiple R-squared: 0.003938, Adjusted R-squared: -0.004575
## F-statistic: 0.4626 on 1 and 117 DF, p-value: 0.4978
# Data Exploration and Analyses for proportion of leaf eaten
descdist(insects6$prop_eaten, discrete = FALSE)
```

Cullen and Frey graph

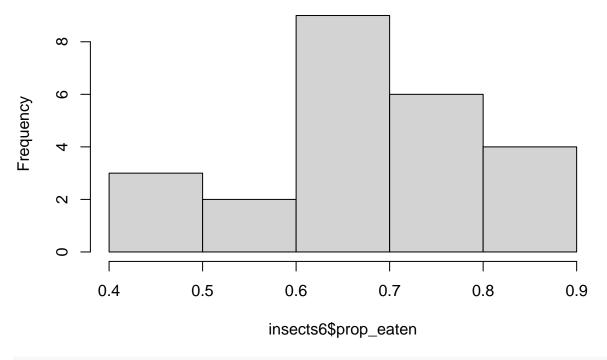


```
## summary statistics
## -----
## min: 0.4126754 max: 0.8605882
## median: 0.6938382
## mean: 0.6716924
## estimated sd: 0.1181775
## estimated skewness: -0.5348535
```

estimated kurtosis: 2.937135

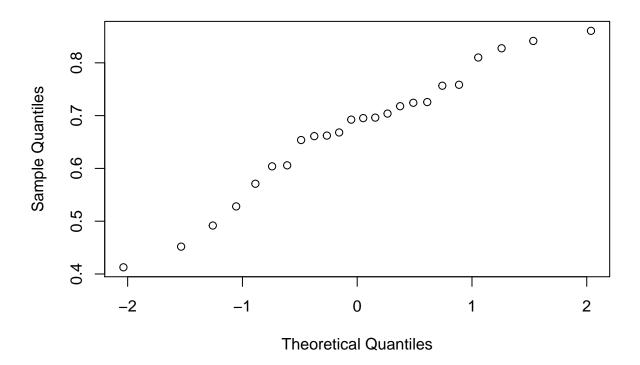
hist(insects6\$prop_eaten)

Histogram of insects6\$prop_eaten



qqnorm(insects6\$prop_eaten)

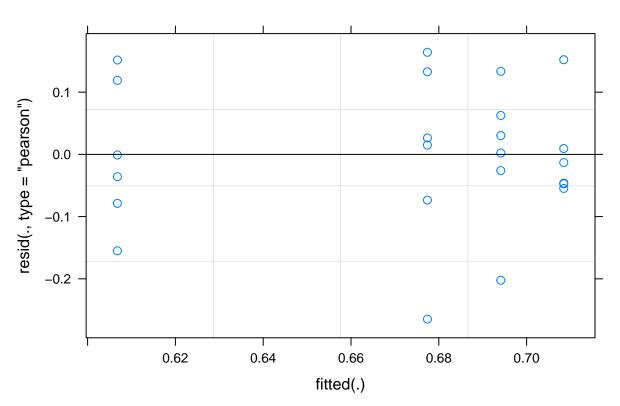
Normal Q-Q Plot



```
shapiro.test(insects6$prop_eaten) # p-value = 0.3238 so we can't reject the null hypothesis that the d
```

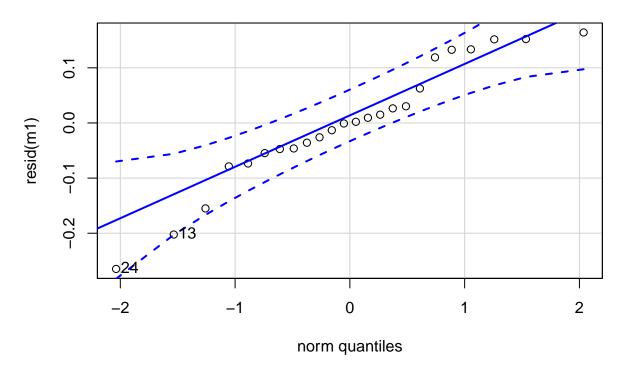
Proportion Eaten

residuals vs predicted values.
plot(m1, main = "Proportion Eaten")



```
# Homogeneity of variance is ok here (increasing variance in # resids is not increasing with fitted values) Check for # homogeneity of variances (true if p>0.05). If the result is
```

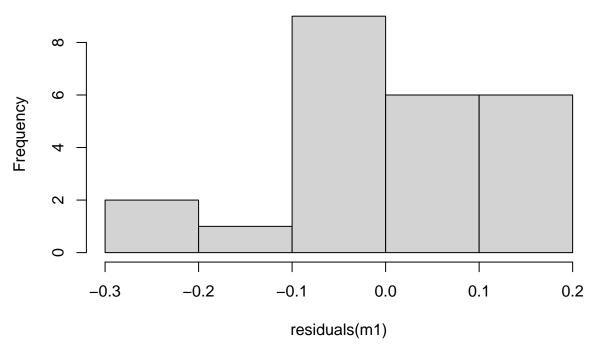
Proportion Eaten



24 13 ## 11 6

hist(residuals(m1), main = "Proportion Eaten")

Proportion Eaten



```
shapiro.test(resid(m1)) # Normal
##
   Shapiro-Wilk normality test
##
##
## data: resid(m1)
## W = 0.94654, p-value = 0.2277
# Model comparisons
m2 <- lm(prop_eaten ~ subplot_descriptions, data = insects6)</pre>
m3 <- lmer(prop_eaten ~ subplot_descriptions + (1 | replicate),</pre>
    data = insects6, REML = F)
## boundary (singular) fit: see ?isSingular
m4 <- lmer(prop_eaten ~ subplot_descriptions + (1 | replicate/footprint_location),</pre>
    data = insects6, REML = FALSE)
## boundary (singular) fit: see ?isSingular
AICctab(m1, m2, m3, m4, weights = T)
      dAICc df weight
## m2 0.0
            5 0.740
```

m1 3.6

m3 3.6

m4 7.7

6 0.122

6 0.122

7 0.016

```
summary(m2)
##
## Call:
## lm(formula = prop_eaten ~ subplot_descriptions, data = insects6)
## Residuals:
##
        Min
                   1Q
                         Median
## -0.264733 -0.049184 0.000584 0.076642 0.164070
##
## Coefficients:
                                      Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                                      0.67741
                                                 0.04870 13.909 9.6e-12 ***
## subplot_descriptionsdrought
                                      -0.07062
                                                 0.06887 -1.025
                                                                     0.317
## subplot_descriptionswarmed
                                      0.03105
                                                            0.451
                                                                     0.657
                                                 0.06887
## subplot_descriptionswarmed_drought 0.01671
                                                 0.06887
                                                            0.243
                                                                     0.811
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Residual standard error: 0.1193 on 20 degrees of freedom
## Multiple R-squared: 0.1139, Adjusted R-squared: -0.01897
## F-statistic: 0.8572 on 3 and 20 DF, p-value: 0.4793
summ(m2)
```

Observations	24
Dependent variable	prop_eaten
Type	OLS linear regression

F(3,20)	0.86
\mathbb{R}^2	0.11
$Adj. R^2$	-0.02

	Est.	S.E.	t val.	p
(Intercept)	0.68	0.05	13.91	0.00
$subplot_descriptions drought$	-0.07	0.07	-1.03	0.32
$subplot_descriptions warmed$	0.03	0.07	0.45	0.66
$subplot_descriptions warmed_drought$	0.02	0.07	0.24	0.81

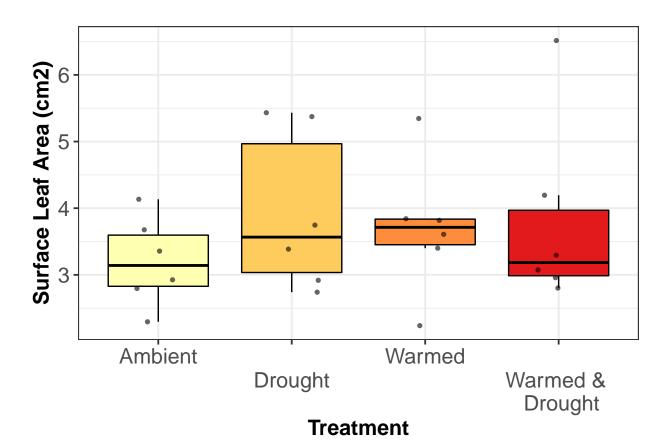
Standard errors: OLS

Model 2 fits the best

```
# Post hoc test to compare different levels
emmeans(m2, list(pairwise ~ subplot_descriptions), adjust = "tukey")
```

```
## $'emmeans of subplot_descriptions'
## subplot_descriptions emmean SE df lower.CL upper.CL
## ambient 0.677 0.0487 20 0.576 0.779
## drought 0.607 0.0487 20 0.505 0.708
```

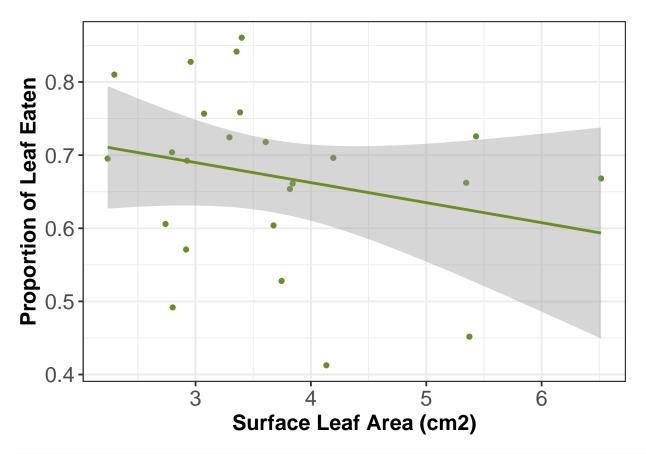
```
## warmed
                          0.708 0.0487 20
                                             0.607
                                                      0.810
## warmed_drought
                         0.694 0.0487 20
                                             0.593
                                                      0.796
##
## Confidence level used: 0.95
##
## $'pairwise differences of subplot descriptions'
                                          SE df t.ratio p.value
## 1
                            estimate
                              0.0706 0.0689 20 1.025 0.7368
## ambient - drought
## ambient - warmed
                              -0.0310 0.0689 20 -0.451 0.9687
## ambient - warmed_drought -0.0167 0.0689 20 -0.243 0.9948
## drought - warmed
                              -0.1017 0.0689 20 -1.476 0.4696
## drought - warmed_drought -0.0873 0.0689 20 -1.268 0.5928
                              0.0143 0.0689 20 0.208 0.9967
## warmed - warmed_drought
##
## P value adjustment: tukey method for comparing a family of 4 estimates
# Looking at SLA (leaf traits) at subplot level
sla_before_trt <- left_join(meta, sla_before)</pre>
## Joining, by = c("replicate", "footprint_location", "subplot_location")
sla_before_trt <- na.omit(sla_before_trt)</pre>
# png('soca_SLA_treatments.png', units='in', width=7,
# height=6, res=300)
ggplot(sla_before_trt, aes(x = subplot_descriptions, y = sla_before_mean,
   fill = subplot_descriptions)) + geom_boxplot(color = "black",
   outlier.shape = NA) + geom_jitter(shape = 16, position = position_jitterdodge(),
    alpha = 0.6) + # geom_errorbar(aes(ymin = lower, ymax =upper)) +
labs(x = "Treatment", y = "Surface Leaf Area (cm2)") + scale_fill_brewer(palette = "Y10rRd") +
    scale_x_discrete(limits = c("ambient", "drought", "warmed",
        "warmed_drought"), labels = c(ambient = "Ambient", drought = "Drought",
       warmed = "Warmed", warmed_drought = "Warmed & \n Drought"),
       guide = guide_axis(n.dodge = 2)) + theme(legend.position = "none")
```



```
# dev.off()
# not much of a difference between treatments in SLA
```

```
# Regression
lm_plot_prop <- lm(prop_eaten ~ sla_before_mean, data = insects6)
# png('regression_preference.png', units='in', width=5,
# height=5, res=300)
ggplot(insects6, aes(sla_before_mean, prop_eaten)) + geom_point(color = "olivedrab") +
        geom_smooth(method = "lm", color = "olivedrab") + labs(x = "Surface Leaf Area (cm2)",
        y = "Proportion of Leaf Eaten")</pre>
```

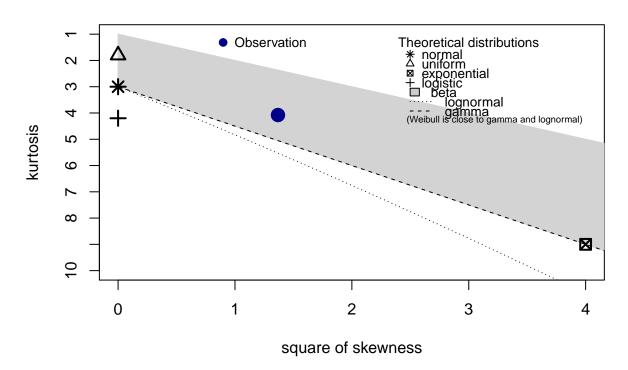
'geom_smooth()' using formula 'y ~ x'



```
# dev.off()
summary(lm_plot_prop)
```

```
##
## Call:
## lm(formula = prop_eaten ~ sla_before_mean, data = insects6)
##
## Residuals:
                  1Q
                      Median
                                           Max
## -0.24607 -0.07330 0.02255 0.07571 0.18174
##
## Coefficients:
                  Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                   0.77198
                              0.08716
                                        8.857 1.05e-08 ***
## sla_before_mean -0.02739
                              0.02289
                                       -1.196
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Residual standard error: 0.1171 on 22 degrees of freedom
## Multiple R-squared: 0.06109, Adjusted R-squared: 0.01842
## F-statistic: 1.432 on 1 and 22 DF, p-value: 0.2443
# Data Exploration and Analyses for SLA
descdist(sla_before_trt$sla_before_mean, discrete = FALSE)
```

Cullen and Frey graph



```
## summary statistics
```

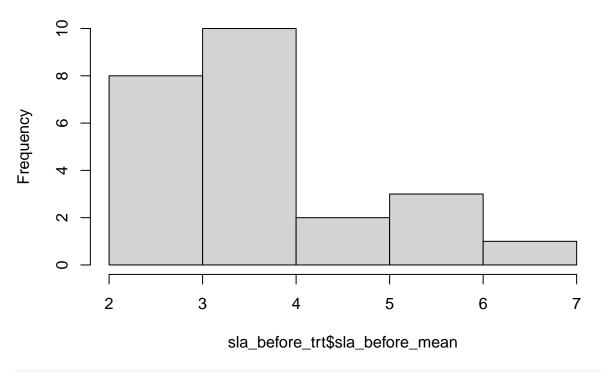
min: 2.238 max: 6.515

median: 3.393 ## mean: 3.661208

estimated sd: 1.066401 ## estimated skewness: 1.16953 ## estimated kurtosis: 4.077587

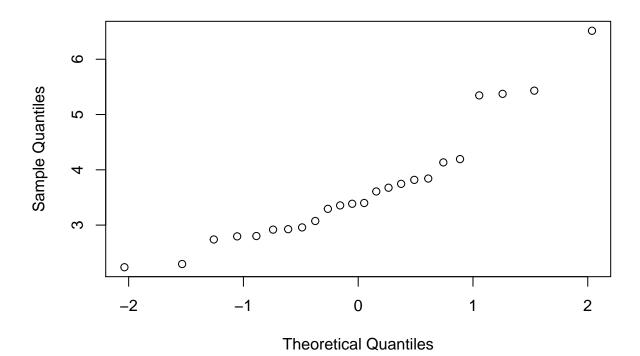
hist(sla_before_trt\$sla_before_mean) # right skewed

Histogram of sla_before_trt\$sla_before_mean



qqnorm(sla_before_trt\$sla_before_mean)

Normal Q-Q Plot



```
shapiro.test(sla_before_trt$sla_before_mean) # p-value = 0.01431 so we reject the null hypothesis that
```

```
##
## Shapiro-Wilk normality test
##
## data: sla_before_trt$sla_before_mean
## W = 0.89155, p-value = 0.01431
```